## stat430 a4 q2

## 2023-11-23

```
x1 = rep(c(-1,1), times=8)
x2 = rep(c(-1,1), times=4, each=2)
x3 = rep(c(-1,1), times=2, each=4)
y = c(5,15,19,11,4,26,-3,21,5,15,21,13,12,17,2,17)
datq2 = data.frame(y,x1,x2,x3)
datq2
##
       y x1 x2 x3
## 1
     5 -1 -1 -1
## 2 15 1 -1 -1
## 3 19 -1 1 -1
## 4 11 1 1 -1
## 5
      4 -1 -1 1
## 6 26 1 -1 1
## 7 -3 -1 1 1
## 8 21 1 1 1
      5 -1 -1 -1
## 9
## 10 15 1 -1 -1
## 11 21 -1 1 -1
## 12 13 1 1 -1
## 13 12 -1 -1 1
## 14 17 1 -1 1
## 15 2 -1 1 1
## 16 17 1 1 1
(a)
# main effect of factor A
y_A_high <- mean(datq2$y[datq2$x1==1])</pre>
y_A_{low} \leftarrow mean(datq2\$y[datq2\$x1==-1])
ME_A <- y_A_high - y_A_low</pre>
\mathtt{ME}_{-}\mathtt{A}
## [1] 8.75
(b)
# Interaction Effect between A and B
x12 < - x1 * x2
data <- data.frame(y,x12)</pre>
y_AB_high <- mean(data$y[data$x12==1])</pre>
y_AB_low <- mean(data$y[data$x12==-1])</pre>
```

```
IE_AB <- y_AB_high - y_AB_low</pre>
IE_AB
## [1] -3
(c)
# Interaction Effect between A,B,C
x123 <- x12 * x3
data <- data.frame(y,x123)</pre>
y_ABC_high <- mean(data$y[data$x123==1])</pre>
y_ABC_low <- mean(data$y[data$x123==-1])</pre>
IE_ABC <- y_ABC_high - y_ABC_low</pre>
IE_ABC
## [1] 6
(d)
# simplest model
mod1 \leftarrow lm(y \sim x1 + x2 + x3, data = datq2)
summary(mod1)
##
## Call:
## lm(formula = y \sim x1 + x2 + x3, data = datq2)
## Residuals:
      Min
             1Q Median
                           3Q
                                  Max
## -10.75 -3.75 -2.25 4.50 12.25
##
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 12.500 1.839 6.796 1.91e-05 ***
## x1
                            1.839
                                    2.379 0.0348 *
                  4.375
## x2
                  0.125
                            1.839
                                    0.068 0.9469
## x3
                 -0.500
                            1.839 -0.272 0.7904
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 7.357 on 12 degrees of freedom
## Multiple R-squared: 0.3234, Adjusted R-squared: 0.1543
## F-statistic: 1.912 on 3 and 12 DF, p-value: 0.1815
beta1 <- summary(mod1)$coefficients[2]</pre>
beta1
## [1] 4.375
# check work from (a)
all.equal(2 * beta1, ME_A)
```

## [1] TRUE

Comments: We found that 2 times estimate of beta1 equals to the estimate of main effect A, we verified our work.

(e)

```
# full model
mod2 \leftarrow lm(y \sim x1 + x2 + x3 + x1:x2 + x1:x2:x3)
summary(mod2)
##
## Call:
## lm(formula = y \sim x1 + x2 + x3 + x1:x2 + x1:x2:x3)
## Residuals:
##
     Min
              1Q Median
                            3Q
                                   Max
## -9.250 -4.438 0.500 3.000 11.250
##
## Coefficients:
##
               Estimate Std. Error t value Pr(>|t|)
                                     7.297 2.61e-05 ***
                 12.500
                             1.713
## (Intercept)
                  4.375
                             1.713
                                     2.554
                                              0.0287 *
## x1
                                     0.073
## x2
                  0.125
                             1.713
                                              0.9433
## x3
                 -0.500
                             1.713 -0.292
                                              0.7763
                 -1.500
                             1.713 -0.876
                                              0.4018
## x1:x2
                  3.000
                             1.713
                                     1.751
                                              0.1104
## x1:x2:x3
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 6.852 on 10 degrees of freedom
## Multiple R-squared: 0.5109, Adjusted R-squared: 0.2664
## F-statistic: 2.089 on 5 and 10 DF, p-value: 0.1504
# check work from (b)
beta4 <- summary(mod2)$coefficients[5]</pre>
beta4
## [1] -1.5
all.equal(2 * beta4, IE_AB)
## [1] TRUE
# check work from (c)
beta5 <- summary(mod2)$coefficients[6]</pre>
beta5
## [1] 3
all.equal(2 * beta5, IE_ABC)
```

## [1] TRUE

Comments: We found that 2 times estimates of beta4 and beta5 equals to the estimates of Interaction Effect between A&B, A&B&C separately, we verified our work.

**(f)** 

```
X <- cbind(x1,x2,x3) # design matrix
XTX <- t(X) %*% X
XTX_inverse <- solve(XTX)
XTX_inverse</pre>
```

```
## x1 0.0625 0.0000 0.0000
## x2 0.0000 0.0625 0.0000
## x3 0.0000 0.0000 0.0625
```

Comments: Obviously, the matrix is disgonal.