## NTHU STAT 5410 - Linear Models Assignment 4 Report

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```
1.
> gala <- read.table("C:/Users/Thomas/Downloads/Linear_mo</pre>
dels/hw4/E3.7.txt", header=T)
> y <- gala[,7]</pre>
> x1 <- gala[,2]
> x2 <- gala[,3]</pre>
> x3 <- gala[,4]</pre>
> x4 <- gala[,5]
> x5 <- gala[,6]</pre>
> fit <- lm(y \sim x1 + x2 + x3 + x4 + x5)
> summary(fit)
call:
lm(formula = y \sim x1 + x2 + x3 + x4 + x5)
Residuals:
                  Median
    Min
             10
                               3Q
                                      Max
-0.39447 -0.11847 0.00053 0.08313
                                      0.56232
Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept) -2.156e+00 9.135e-01 -2.360
                                             0.0333 *
           -9.012e-06 5.184e-04 -0.017 0.9864
x1
x2
           1.316e-03 1.263e-03 1.041 0.3153
x3
           1.278e-04 7.690e-05
                                   1.662 0.1188
           7.899e-03 1.400e-02
x4
                                   0.564
                                          0.5815
           1.417e-04 7.375e-05
x5
                                   1.921
                                           0.0754 .
>cv <- qt(0.975, fit$df)
# the 95% critical value, dfw=n-p=14
```

```
(a)
> c(-9.012e-06 - cv * 5.184e-04, -9.012e-06 + cv * 5.184e-
04)
[1] -0.001120869 0.001102845
The 95% C.I. for \beta_1 is [-0.001120869, 0.001102845].
(b)
> A \leftarrow t(c(0, 0, 0, 1, 0, 2))
> y0 <- sum(A * fit$coef)</pre>
> y0
[1] 0.0004111073
> x <- model.matrix(fit) # the model matrix X
> xtxi <- solve(t(x)%*%x) # (XTX)-1
> bm <- sqrt(A%*%xtxi%*%t(A)) * summary(fit)$sigma</pre>
> bm
            [,1]
[1,] 0.0001641751
> cv <- qt(0.975, fit$df)
> c(y0 - cv * bm, y0 + cv * bm)
[1] 5.898666e-05 7.632279e-04
The 95% C.I. for \beta_3 + 2\beta_5 is [0.00005898666, 0.0007632279].
2.
> gala <- read.table("C:/Users/Thomas/Downloads/Linear_mo</pre>
dels/hw4/set.txt", header=T)
> PRICE <- gala[,1]</pre>
> BDR <- gala[,2]
> FLR <- gala[,3]
> FP <- gala[,4]
> RMS <- gala[,5]
> ST <- gala[,6]
> LOT <- gala[,7]
> TAX <- gala[,8]
> BTH <- gala[,9]
> CON <- gala[,10]
> GAR <- gala[,11]
```

```
> CDN <- gala[,12]
> L1 <- gala[,13]
> L2 <- gala[,14]</pre>
```

Since the new data we try to predict does not contain all the factors in the data set, we only use the factors that the new data has to fit our model.

```
> fit <-
lm(PRICE ~ BDR + FLR + FP + RMS + ST + LOT + BTH + GAR)
> summary(fit)

Call:
lm(formula =
PRICE ~ BDR + FLR + FP + RMS + ST + LOT + BTH + GAR)

Residuals:
    Min     1Q     Median     3Q     Max
-10.3058     -2.8417     -0.1511     3.2882     7.9518
```

## Coefficients:

```
Estimate Std. Error t value Pr(>|t|)
                       5.240957
                                 3.556 0.002429 **
(Intercept) 18.637664
BDR
          -7.697444
                     1.829426 -4.208 0.000592 ***
                     0.003235 5.431 4.49e-05 ***
FLR
           0.017570
                     3.083583
                               2.241 0.038680 *
FΡ
           6.909765
           3.904374
                     1.615617 2.417 0.027194 *
RMS
ST
          10.818663 2.300203
                              4.703 0.000205 ***
           0.263522 0.135109
                               1.950 0.067808 .
LOT
BTH
           2.374591
                     2.557865
                               0.928 0.366221
           1.770861
                     1.404310
                               1.261 0.224334
GAR
```

The fitted model is

```
PRICE = 18.637664 - 7.697444 *BDR + 0.017570 *FLR + 6.909765 *FP + 3.904374 *RMS + 10.818663 *ST + 0.263522 *LOT + 2.374591 *BTH + 1.770861 *GAR
```

To price we try to predict for the new data, we first check if it is an interpol ation or extrapolation. Since all the new predicators lies in the range of the pr edictors in the data set, it is an interpolation.

## > summary(gala)

```
Price
               BDR
                          FLR
                                                 RMS
                                                            ST
                                                                       LOT
                                         :0.0000
1st Qu.:46.25    1st Qu.:2.000    1st Qu.: 806    1st Qu.:0.0000    1st Qu.: 5.0    1st Qu.:0.0000    1st Qu.:25.50
Median :55.50 Median :3.000 Median : 987
                                    Median: 0.0000 Median: 6.0 Median: 0.0000 Median: 30.00
                                        :0.1538 Mean : 6.5 Mean :0.2692 Mean
Mean :56.15 Mean :3.231 Mean :1100 Mean
3rd Qu.:64.00 3rd Qu.:4.000 3rd Qu.:1204 3rd Qu.:0.0000 3rd Qu.: 7.0 3rd Qu.:0.7500 3rd Qu.:36.50
                                         :1.0000
               втн
                                                 CDN
                                                                         L2
                          CON
                                    GAR
                                                             L1
Min. : 440.0 Min. :1.000 Min.
                             :0.0
                                   Min.
                                        :0.0000 Min. :0.0000 Min.
                                                                  :0.0000 Min. :0.0000
Median: 817.0 Median: 1.500 Median: 0.5 Median: 1.0000 Median: 0.0000 Median: 0.0000 Median: 0.0000
Mean : 898.1 Mean :1.481 Mean
                             :0.5 Mean
                                        :0.8462 Mean
                                                     :0.2308 Mean
3rd Qu.: 991.0 3rd Qu.:1.875 3rd Qu.:1.0 3rd Qu.:1.5000 3rd Qu.:0.0000 3rd Qu.:1.0000 3rd Qu.:1.0000
    :2700.0 Max. :3.000 Max. :1.0 Max. :2.0000 Max.
                                                     :1.0000 Max.
```

To predict a future observation, we use

> predict(fit, data.frame(BDR=2, FLR=750, FP=1, RMS=5, ST=
1, LOT=25, BTH=1.5, GAR=1), se=T, interval="prediction")
\$fit

fit lwr upr 1 65.59152 52.89018 78.29287

\$se.fit
[1] 3.740865

\$df [1] 17

\$residual.scale
[1] 4.716756

The predicted price is 65.59152.

The 95% C.I. for predicted price is [52.89018, 78.29287].