

NTHU STAT 5410 - Linear Models

Assignment 4 Report

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1.

```
> gala <- read.table("C:/Users/Thomas/Downloads/Linear_models/hw4/E3.7.txt", header=T)
> y <- gala[,7]
> x1 <- gala[,2]
> x2 <- gala[,3]
> x3 <- gala[,4]
> x4 <- gala[,5]
> x5 <- gala[,6]
> fit <- lm(y ~ x1 + x2 + x3 + x4 + x5)
> summary(fit)
```

Call:

```
lm(formula = y ~ x1 + x2 + x3 + x4 + x5)
```

Residuals:

Min	1Q	Median	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 *
x1	-9.012e-06	5.184e-04	-0.017	0.9864
x2	1.316e-03	1.263e-03	1.041	0.3153
x3	1.278e-04	7.690e-05	1.662	0.1188
x4	7.899e-03	1.400e-02	0.564	0.5815
x5	1.417e-04	7.375e-05	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 β_1 is [-0.001120869, 0.001102845].

(b)

```
> A <- t(c(0, 0, 0, 1, 0, 2))
> y0 <- sum(A * fit$coef)
> y0
[1] 0.0004111073
> x <- model.matrix(fit) # the model matrix X
> xtxi <- solve(t(x)%*%x) # (XTX)-1
> bm <- sqrt(A%*%xtxi%*%t(A)) * summary(fit)$sigma
> 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_models/hw4/set.txt", header=T)
> PRICE <- gala[,1]
> 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]
```

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

The fitted model is

$$\begin{aligned}
 PRICE = & 18.637664 - 7.697444 * BDR + 0.017570 * FLR + 6.909765 * FP \\
 & + 3.904374 * RMS + 10.818663 * ST + 0.263522 * LOT \\
 & + 2.374591 * BTH + 1.770861 * GAR
 \end{aligned}$$

The price we try to predict for the new data

```
> 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 the predicted price is [52.89018 78.29287].