

151AHW2

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5-(c)

From 5.2.2, we know that

Prestige = -6.7943 + 4.1866 Education + 0.0013136 Income - 0.0089052 Woman

$\sigma_{\hat{\epsilon}} = 7.846$

degrees of freedom for t-distribution : $102 - 3 - 1 = 98$ (n-p-1)

$x_0^T = (1, 13, 12000, 50)$.

```
data <- matrix(0, nrow=4, ncol=4)
colnames(data)<- c("Pres", "Educ", "Inc", "%W")
rownames(data)<- c("Pres", "Educ", "Inc", "%W")

uppertri <- c(253618, 55326, 12513, 37748108, 8121410,
6534383460, 131909, 32281, 14093097, 187312)

data[upper.tri(data, diag=TRUE)] <- uppertri
data[lower.tri(data)] <- t(data)[lower.tri(data)]

finalsum <- matrix(c(102, 1095, 693386, 2956), nrow=4, ncol = 1)
# we don't need prestige, replace sum of prestige as n = 102 (sum of 1*102)

XtX <- cbind(finalsum, rbind(finalsum[-1],data[-1,-1]))
XtXinv <- solve(XtX)

t_0.05 <- qt(p=0.95, df=102-3-1)
beta <- c(-6.7943, 4.1866, 0.0013136, -0.0089052)
x_0 <- c(1, 13, 12000, 50)
std_err <- 7.846

var_delta <- std_err^2 * x_0 %*% XtXinv %*% x_0      # var(Y_hat0 - E(Y0))
var_D <- std_err^2 * (1 + x_0 %*% XtXinv %*% x_0 )  # var(Y_hat0 - Y0)

cat("delta = Y_hat0 - E(Y0)", "D = Y_hat0 - Y0\n", sep = "\n")

## delta = Y_hat0 - E(Y0)
## D = Y_hat0 - Y0

cat("var(delta) : ", var_delta, "\nvar(D): ", var_D, "\n\n")

## var(delta) : 2.754159
## var(D): 64.31387
```

```

Y_hat0 <- x_0 %*% beta;
cat("Y_hat0 (Point estimate) : \n")

## Y_hat0 (Point estimate) :
as.numeric(Y_hat0)

## [1] 62.94944
cat("Confidence interval for E(Y_0)\n")

## Confidence interval for E(Y_0)
Y_hat0 + c(-1,1) * t_0.05 * sqrt(var_delta)

## [1] 60.19365 65.70523
cat("Confidence interval for Y_0\n")

## Confidence interval for Y_0
Y_hat0 + c(-1,1) * t_0.05 * sqrt(var_D)

## [1] 49.63249 76.26639

```

5 - (d)

```

x_0 <- c(1, 0, 50000, 100)
Y_hat <- as.numeric(x_0 %*% beta)
cat("Y_hat (Estimated point) :", Y_hat, "\n")

## Y_hat (Estimated point) : 57.99518

var_delta <- std_err^2 * x_0 %*% XtXinv %*% x_0
var_D <- std_err^2 * (1 + x_0 %*% XtXinv %*% x_0 )
cat("var(delta) : ", var_delta, "\nvar(D): ", var_D)

## var(delta) : 271.7183
## var(D): 333.278

```

Estimated variance of the forecast error ($\text{Var}(D) = \text{Var}(Y_{\text{hat}} - Y)$) is 64.3 in part (c) while it is 333.278 in part(d). Since $\text{Var}(D)$ in part(d) is almost 5 times greater than $\text{Var}(D)$ in part(c), it suggests that error is very large and so does uncertainty. It is understandable since there is no occupation which has similar characteristics as the given data.

(given data : an occupation with an average income of \$50,000, an average education of 0 years, and 100% women.)

6-(a)

```

dat <- read.csv("/Users/cloverjiyoon/2017Fall/Stat 151A/Lab/Lab3/bodyfat.csv")
n = dim(dat)[1]
p <- 4
q <- 1

fit <- lm(bodyfat ~ Knee + Thigh + Hip + Ankle, data = dat)
RSS_M <- sum(resid(fit)^2)

```

$$H_0 : \beta_{knee} + \beta_{thigh} = \beta_{hip} + \beta_{ankle}$$

Since $\beta_{knee} = -\beta_{thigh} + \beta_{hip} + \beta_{ankle}$, the model can then be rewritten as

$$\text{bodyfat} = \beta_0 + \beta_{thigh}(\text{thigh} - \text{knee}) + \beta_{hip}(\text{hip} + \text{knee}) + \beta_{ankle}(\text{ankle} + \text{knee})$$

Use formula

$$\frac{(\text{RSS}(m) - \text{RSS}(M))/q}{\text{RSS}(M)/(n - p - 1)}$$

where q is the number of dropped variable.

```
fit_m <- lm(bodyfat ~ I(Thigh - Knee) + I(Hip + Knee) + I(Ankle + Knee), data=dat)
RSS_m <- sum(resid(fit_m)^2)
Fstat1 <- ((RSS_m - RSS_M) / q) / (RSS_M / (n - p - 1))
```

```
cat("F statistics is : ", Fstat1, "\n")
```

```
## F statistics is : 0.002813842
```

```
cat("Degree of freedom is (1, 247)\n")
```

```
## Degree of freedom is (1, 247)
```

```
cat("P value is : ", 1- pf(Fstat1,1,n-p-1), "\n")
```

```
## P value is : 0.9577384
```

General way to solve problem =====

Can write hypothesis as $H_0 : L\beta = 0$, where

$$L = \begin{bmatrix} 0 & 1 & 1 & -1 & -1 \end{bmatrix}$$

Use formula

$$\frac{(L\hat{\beta} - c)^T [L(X^T X)^{-1} L^T]^{-1} (L\hat{\beta} - c) / q}{\text{RSS}(M)/(n - p - 1)}.$$

```
X <- as.matrix(cbind(1, dat[,c("Knee", "Thigh", "Hip", "Ankle")]))
y <- as.numeric(dat$bodyfat)
beta_hat <- solve(t(X) %*% X) %*% t(X) %*% y
y_hat <- X %*% beta_hat
L <- matrix(c(0,1,1,-1,-1), nrow = 1)
Fstat2 <- (t(L %*% beta_hat) %*% solve(L %*% solve(t(X) %*% X) %*% t(L)) %*% (L %*% beta_hat) / q) / (RSS_M / (n - p - 1))
```

```
cat("F statistics is : ", Fstat2, "\n")
```

```
## F statistics is : 0.002813842
```

```
cat("Degree of freedom is (1, 247)\n")
```

```
## Degree of freedom is (1, 247)
```

```
cat("P value is : ", 1- pf(Fstat2,1,n-p-1) , "\n")
```

```
## P value is : 0.9577384
```

Check that they match.

```
Fstat1
```

```
## [1] 0.002813842
```

```
Fstat2
```

```
##           [,1]  
## [1,] 0.002813842
```

Since P value is fairly large, we do not reject the null hypothesis.

6-(b)

```
XtXinv <- solve(t(X) %*% X)  
S <- sqrt(sum(resid(fit)^2)/(n-p-1)) # RSE = sqrt(RSS/n-p-1)  
var_mat <- S^2 * XtXinv  
numerator <- beta_hat[2] + beta_hat[3] - beta_hat[4] - beta_hat[5]  
denominator <- sqrt(sum(diag(var_mat)[2:5]) + 2*var_mat[2,3] - 2*var_mat[2,4] - 2*var_mat[2,5] - 2*var_mat[3,4] + 2*var_mat[3,5] - 2*var_mat[4,5])
```

```
t <- as.numeric(numerator / denominator)  
cat("T statistics is: ", t, "\n")
```

```
## T statistics is: 0.05304566
```

```
cat("Degree of Freedom : ", n-p-1, "\n")
```

```
## Degree of Freedom : 247
```

```
cat("P value : ", 2*(1-pt(t,247)), "\n")
```

```
## P value : 0.9577384
```

6-(c)

```
cat("Square of T statistics : ", t^2, "\n")
```

```
## Square of T statistics : 0.002813842
```

```
cat("F statistics : ", Fstat1, "\n")
```

```
## F statistics : 0.002813842
```

As we can see, square value of T statistics is equal to the F statistics value.