

# HW 3 Problem 1

$$\beta_{12} = \beta_{33}$$

$$\beta_{22}\beta_{32} = 0$$

1) The null hypothesis is there is no mediation effect of x on y through m.

2) From code:

```
setwd("/Users/alwinlin/Desktop/DSC 382 Regression")
data <- read.csv(file = 'HW3_data.csv', header = TRUE)
eq1 <- lm(y ~ x, data = data) #First equation
eq2 <- lm(m ~ x, data = data) #Second equation regress m on y
eq3 <- lm(y ~ m + x, data = data) #Third equation regress the last equation with m & x
beta11 <- coef(eq1)[1]
beta12 <- coef(eq1)[2]
beta21 <- coef(eq2)[1]
beta22 <- coef(eq2)[2]
beta31 <- coef(eq3)[1]
beta32 <- coef(eq3)[2]
beta33 <- coef(eq3)[3]
```

$$\begin{matrix} \beta_{11} = -1.61 & \beta_{12} = 2.64 & \beta_{21} = -0.712 \\ \beta_{22} = 2.06 & \beta_{31} = -0.666 \\ \beta_{32} = 1.22 & \beta_{33} = 0.133 \end{matrix}$$

3) knowing  $\sigma^2 = 1$  &  $\text{var} = \sigma^2 C_j$  (slide 24)  
 $C_j = (X'X)^{-1}$  and var is the second diag

```
Matrix2 <- model.matrix(eq2)
vareq2 <- solve(t(Matrix2) %*% Matrix2)
var22 <- diag(vareq2)[2] # Variance of beta22
Matrix3 <- model.matrix(eq3)
vareq3 <- solve(t(Matrix3) %*% Matrix3)
var32 <- diag(vareq3)[2] # Variance of beta32
```

$$\begin{matrix} \text{var}(\beta_{22}) = 0.00113 \\ \text{var}(\beta_{32}) = 0.0084 \end{matrix}$$

$$4) Z = \frac{\hat{\beta}_{12} - \hat{\beta}_{33}}{\sqrt{\hat{\beta}_{22}^2 \text{var}(\hat{\beta}_{32}) + \hat{\beta}_{32}^2 \text{var}(\hat{\beta}_{22})}} =$$

```
> Z_sobel
      x
12.99453
```

$$Z = 12.995$$

```
Z_sobel <- (beta12 - beta33) / sqrt((beta22^2 * var32) + (beta32^2 * var22))
```

5)  $p = 2(1 - \Phi(|Z|)) \approx 0$  We reject the null hypothesis at the 5% significance level of no mediation effect

```
p_val <- 2 * (1 - pnorm(abs(Z_sobel)))
```

p_val	Named num 0
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## Problem 2

1) residual sum of squares =  $\sum_{i=1}^n (y_i - x_i \beta)^2$ ,  $\frac{d}{d\beta} (RSS) = -2 \sum_{i=1}^n x_i (y_i - x_i \beta) = 0$

$\sum_{i=1}^n x_i y_i - \beta \sum_{i=1}^n x_i^2 = 0 \rightarrow \hat{\beta} = \frac{\sum_{i=1}^n x_i y_i}{\sum_{i=1}^n x_i^2}$  Since  $\sum_{i=1}^n x_i^2 = n$   $\hat{\beta} = \frac{\sum_{i=1}^n x_i y_i}{n} = n^{-1} \sum_{i=1}^n x_i y_i$

2) RSS from above =  $\sum_{i=1}^n (y_i - x_i \beta)^2$  expand  $\rightarrow \sum_{i=1}^n y_i^2 - 2\beta \sum_{i=1}^n x_i y_i + \beta^2 \sum_{i=1}^n x_i^2$

We know  $\sum_{i=1}^n x_i^2 = n$ , so  $\sum_{i=1}^n x_i y_i = n\hat{\beta}$ . Substitute into  $\uparrow$  and get:

$RSS = \sum_{i=1}^n y_i^2 - 2\hat{\beta}(n\hat{\beta}) + \hat{\beta}^2(n) = \boxed{\sum_{i=1}^n y_i^2 - n\hat{\beta}^2}$

3)  $F = \frac{\hat{e}'_{red} \hat{e}_{red} - \hat{e}' \hat{e}}{\hat{e}' \hat{e} / (n-p)}$  where  $\hat{e}'_{red} \hat{e}_{red} = \sum_{i=1}^n y_i^2$ ,  $\hat{e}' \hat{e} = \sum_{i=1}^n y_i^2 - n\hat{\beta}^2$   
 $p=1$

numerator =  $\hat{e}'_{red} \hat{e}_{red} - \hat{e}' \hat{e} = \sum_{i=1}^n y_i^2 - (\sum_{i=1}^n y_i^2 - n\hat{\beta}^2) = n\hat{\beta}^2$

denominator =  $\hat{e}' \hat{e} / (n-p) = \frac{\sum_{i=1}^n y_i^2 - n\hat{\beta}^2}{n-1}$

$F = \frac{n\hat{\beta}^2}{\frac{\sum_{i=1}^n y_i^2 - n\hat{\beta}^2}{n-1}} = \boxed{\frac{(n-1)n\hat{\beta}^2}{\sum_{i=1}^n y_i^2 - n\hat{\beta}^2}} \quad F \sim F_{1, n-1} \text{ under } H_0$

4)  $\hat{\beta} = n^{-1} \sum_{i=1}^n x_i y_i \leftarrow y_i = \sigma \varepsilon_i$  so  $\hat{\beta} = \frac{\sigma}{n} \sum_{i=1}^n x_i \varepsilon_i$

Mean:  $E[\hat{\beta}] = \frac{\sigma}{n} \sum_{i=1}^n x_i E[\varepsilon_i] = 0$

Var:  $\text{Var}(\hat{\beta}) = \frac{\sigma^2}{n^2} \sum_{i=1}^n x_i^2 \leftarrow \sum_{i=1}^n x_i^2 = n \rightarrow \text{Var}(\hat{\beta}) = \frac{\sigma^2}{n}$

$\hat{\beta} | H_0 \sim N(0, \frac{\sigma^2}{n})$

5) From slide 40,  $T = \frac{\hat{\beta}}{\hat{\sigma} \sqrt{1/n}} = \frac{\sqrt{n} \hat{\beta}}{\hat{\sigma}}$   $T^2 = \frac{n\hat{\beta}^2}{\hat{\sigma}^2} = \frac{n\hat{\beta}^2}{\frac{\sum_{i=1}^n (y_i - \hat{y}_i)^2}{(n-1)}} = \boxed{\frac{n\hat{\beta}^2(n-1)}{\sum_{i=1}^n y_i^2 - n\hat{\beta}^2}}$   
boxed is same eq as F from Q3  
 $\sum_{i=1}^n (y_i - \hat{y}_i)^2 = \sum_{i=1}^n y_i^2 - n\hat{\beta}^2$