# Homework 5 Solution

# Question 1

(1)

$$Z^{\top}Z = \begin{bmatrix} 7 & 0 & 0 \\ 0 & 26 & 0 \\ 0 & 0 & 12 \end{bmatrix}$$

$$Z^{\top}Y = \begin{bmatrix} 0 \\ 8 \\ 0 \end{bmatrix}$$

Therefore,

$$\hat{\beta} = (Z^{\top}Z)^{-1}Z^{\top}Y$$

$$= \begin{bmatrix} \frac{1}{7} & 0 & 0\\ 0 & \frac{1}{26} & 0\\ 0 & 0 & \frac{1}{12} \end{bmatrix} \begin{bmatrix} 0\\ 8\\ 0 \end{bmatrix}$$

$$= \begin{bmatrix} 0\\ \frac{4}{13}\\ 0 \end{bmatrix}$$

(2)

$$R^{2} = 1 - \frac{\|\hat{\epsilon}\|^{2}}{\sum_{i=1}^{n} (Y_{i} - \bar{Y})^{2}}$$
$$= 1 - \frac{\|Y - Z\hat{\beta}\|^{2}}{\sum_{i=1}^{n} (Y_{i} - \bar{Y})^{2}}$$
$$= 0.6154$$

(3)

$$\hat{\sigma}^2 = \frac{1}{n - r - 1} \|\hat{\epsilon}\|^2$$
  
= 0.3846

$$\hat{Cov}(\hat{\beta}) = \hat{\sigma}^2 (Z^\top Z)^{-1}$$

$$= \begin{bmatrix} 0.0549 & 0 & 0 \\ 0 & 0.0148 & 0 \\ 0 & 0 & 0.0321 \end{bmatrix}$$

(4) The 95% confidence interval for  $\beta_1$  is

$$[\hat{\beta}_{1} - \hat{\sigma}\sqrt{\omega_{11}}t_{n-r-1}(\frac{0.05}{2}), \hat{\beta}_{1} + \hat{\sigma}\sqrt{\omega_{11}}t_{n-r-1}(\frac{0.05}{2})]$$

$$= [\frac{4}{13} - \sqrt{0.3846}\sqrt{\frac{1}{26}}t_{7-2-1}(\frac{0.05}{2}), \frac{4}{13} + \sqrt{0.3846}\sqrt{\frac{1}{26}}t_{7-2-1}(\frac{0.05}{2})]$$

$$= [-0.0300, 0.6454]$$

(5) The 95% confidence intervals for  $\beta_j,\ j=0,1,2$  based on confidence region are

$$[\hat{\beta}_j - \hat{\sigma}\sqrt{\omega_{jj}}\sqrt{(r+1)F_{r+1,n-r-1}(0.05)}, \hat{\beta}_j + \hat{\sigma}\sqrt{\omega_{jj}}\sqrt{(r+1)F_{r+1,n-r-1}(0.05)}]$$

The results are

$$\beta_0 \in [-1.0423, 1.0423]$$
  
 $\beta_1 \in [-0.2332, 0.8485]$   
 $\beta_2 \in [-0.7961, 0.7961]$ 

(6) The 95% confidence intervals for  $\beta_j,\ j=0,1,2$  based on Bonferroni correction are

$$[\hat{\beta}_j - \hat{\sigma}\sqrt{\omega_{jj}}t_{n-r-1}(\frac{0.05}{2(r+1)}), \hat{\beta}_j + \hat{\sigma}\sqrt{\omega_{jj}}t_{n-r-1}(\frac{0.05}{2(r+1)})]$$

The results are

$$\beta_0 \in [-0.9284, 0.9284]$$
  
$$\beta_1 \in [-0.1740, 0.7894]$$
  
$$\beta_2 \in [-0.7091, 0.7091]$$

(7) Let  $C = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$ . The F-test statistic is

$$\frac{1}{\hat{\sigma}^2} (C\hat{\beta})^{\top} (C(Z^{\top}Z)^{-1}C^{\top})^{-1} (C\hat{\beta}) 
= \frac{1}{\hat{\sigma}^2} \hat{\beta}_{(2)}^T \Omega_{22}^{-1} \hat{\beta}_{(2)} 
= \frac{1}{0.3846} \begin{bmatrix} \frac{4}{13} & 0 \end{bmatrix} \begin{bmatrix} \frac{1}{26} & 0 \\ 0 & \frac{1}{12} \end{bmatrix}^{-1} \begin{bmatrix} \frac{4}{13} \\ 0 \end{bmatrix} 
= 6.4$$

The critical value is

$$(r-q)F_{r-q,n-r-1}(\alpha) = (2-0)F_{2-0,7-2-1}(0.05)$$
  
= 13.8885

As 6.4 < 13.8885, we do not reject the  $H_0$ .

(8)

$$\bar{z}_1 = 0$$
$$\bar{z}_2 = 0$$

So

$$\vec{z}_0 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$$

The confidence interval for the mean response  $E(Y_0)$  given  $\vec{z}_0$  is given by

$$[\vec{z}_0^{\top} \hat{\beta} - \hat{\sigma} t_{n-r-1} (\frac{\alpha}{2}) \sqrt{\vec{z}_0^{\top} (Z^{\top} Z)^{-1} \vec{z}_0}, \vec{z}_0^{\top} \hat{\beta} + \hat{\sigma} t_{n-r-1} (\frac{\alpha}{2}) \sqrt{\vec{z}_0^{\top} (Z^{\top} Z)^{-1} \vec{z}_0}]$$

The result is

$$[-0.6508, 0.6508]$$

(9) The prediction interval for the  $Y_0$  given  $\vec{z}_0$  is given by

$$[\vec{z}_0^{\top} \hat{\beta} - \hat{\sigma} t_{n-r-1} (\frac{\alpha}{2}) \sqrt{1 + \vec{z}_0^{\top} (Z^{\top} Z)^{-1} \vec{z}_0}, \vec{z}_0^{\top} \hat{\beta} + \hat{\sigma} t_{n-r-1} (\frac{\alpha}{2}) \sqrt{1 + \vec{z}_0^{\top} (Z^{\top} Z)^{-1} \vec{z}_0}]$$

The result is

$$[-1.8408, 1.8408]$$

### Question 2

(1)

$$\begin{split} H^\top &= (Z(Z^\top Z)^{-1}Z^\top)^\top \\ &= (Z^\top)^\top ((Z^\top Z)^{-1})^\top Z^\top \\ &= Z((Z^\top Z)^\top)^{-1}Z^\top \\ &= Z(Z^\top Z)^{-1}Z^\top \\ &= H \end{split}$$

$$(I - H)^{\top} = I^{\top} - H^{\top}$$
$$= I - H$$

$$\begin{split} H^2 &= (Z(Z^\top Z)^{-1}Z^\top)(Z(Z^\top Z)^{-1}Z^\top) \\ &= Z(Z^\top Z)^{-1}(Z^\top Z)(Z^\top Z)^{-1}Z^\top \\ &= Z(Z^\top Z)^{-1}Z^\top \\ &= H \end{split}$$

$$(I - H)^2 = (I - H)(I - H)$$
  
=  $I - H - H + H^2$   
=  $I - 2H + H$   
=  $I - H$ 

$$H(I - H) = H - H^2$$
$$= H - H$$
$$= 0$$

(3) Suppose the spectral decomposition of H is

$$H = P\Lambda P^{\top}$$

where P is an orthogonal matrix and  $\Lambda$  is a diagonal matrix. By  $H^2 = H$ , we have

$$\begin{split} P\Lambda P^{\top}P\Lambda P^{\top} &= P\Lambda P^{\top} \\ P\Lambda I\Lambda P^{\top} &= P\Lambda P^{\top} \\ P\Lambda^2 P^{\top} &= P\Lambda P^{\top} \\ P^{\top}(P\Lambda^2 P^{\top})P &= P^{\top}(P\Lambda P^{\top})P \\ \Lambda^2 &= \Lambda \end{split}$$

Because  $\Lambda$  is a diagonal matrix, by  $\Lambda^2 = \Lambda$ , all the eigenvalues of H must be 0 or 1.

$$I - H = PP^{\top} - P\Lambda P^{\top}$$
$$= P(I - \Lambda)P^{\top}$$

So all the eigenvalues of (I - H) must be 0 or 1.

(4) As all the eigenvalues of H are nonnegative, by the definition of positive semi-definite, H is a positive semi-definite matrix. Similarly, (I - H) is also a positive semi-definite matrix.

(5)

$$HZ = (Z(Z^{\top}Z)^{-1}Z^{\top})Z$$
$$= Z(Z^{\top}Z)^{-1}(Z^{\top}Z)$$
$$= Z$$

$$(I - H)Z = IZ - HZ$$
$$= Z - Z$$
$$= 0$$

**Question 3** From Question 2 (5), we know HZ = Z. Then

$$H \begin{bmatrix} Z_{(1)} & Z_{(2)} \end{bmatrix} = \begin{bmatrix} Z_{(1)} & Z_{(2)} \end{bmatrix} \\ \begin{bmatrix} HZ_{(1)} & HZ_{(2)} \end{bmatrix} = \begin{bmatrix} Z_{(1)} & Z_{(2)} \end{bmatrix}$$

Therefore,  $HZ_{(1)} = Z_{(1)}$ .

$$\begin{split} HH_{(red)} &= HZ_{(1)}(Z_{(1)}^{\intercal}Z_{(1)})^{-1}Z_{(1)}^{\intercal} \\ &= (HZ_{(1)})(Z_{(1)}^{\intercal}Z_{(1)})^{-1}Z_{(1)}^{\intercal} \\ &= Z_{(1)}(Z_{(1)}^{\intercal}Z_{(1)})^{-1}Z_{(1)}^{\intercal} \\ &= H_{(red)} \end{split}$$

$$\begin{split} H_{(red)}H &= Z_{(1)}(Z_{(1)}^{\top}Z_{(1)})^{-1}Z_{(1)}^{\top}H \\ &= Z_{(1)}(Z_{(1)}^{\top}Z_{(1)})^{-1}(H^{\top}Z_{(1)})^{\top} \\ &= Z_{(1)}(Z_{(1)}^{\top}Z_{(1)})^{-1}(HZ_{(1)})^{\top} \\ &= Z_{(1)}(Z_{(1)}^{\top}Z_{(1)})^{-1}Z_{(1)}^{\top} \\ &= H_{(red)} \end{split}$$

Here we also use  $H^{\top} = H$  from Question 2 (1).

# Question 4

(1)

$$\begin{split} W &= \begin{bmatrix} 1 & \vec{w}_1^\top \\ 1 & \vec{w}_2^\top \\ \vdots & \ddots \\ 1 & \vec{w}_n^\top \end{bmatrix} \\ &= \begin{bmatrix} 1 & (C\vec{z}_1)^\top \\ 1 & (C\vec{z}_2)^\top \\ \vdots & \ddots \\ 1 & (C\vec{z}_n)^\top \end{bmatrix} \\ &= \begin{bmatrix} 1 & \vec{z}_1^\top \\ 1 & \vec{z}_2^\top \\ \vdots & \ddots \\ 1 & \vec{z}_n^\top \end{bmatrix} \begin{bmatrix} 1 & \vec{0}^\top \\ \vec{0} & C^\top \end{bmatrix} \\ &= Z\tilde{C}^\top \end{split}$$

(2)

$$\begin{split} \hat{\gamma} &= (W^\top W)^{-1} W^\top Y \\ &= (\tilde{C} Z^\top Z \tilde{C}^\top)^{-1} \tilde{C} Z^\top Y \\ &= (\tilde{C}^\top)^{-1} (Z^\top Z)^{-1} \tilde{C}^{-1} \tilde{C} Z^\top Y \\ &= (\tilde{C}^\top)^{-1} (Z^\top Z)^{-1} Z^\top Y \\ &= (\tilde{C}^\top)^{-1} \hat{\beta} \end{split}$$

$$\hat{\epsilon}_w = Y - W\hat{\gamma}$$

$$= Y - Z\tilde{C}^{\top}(\tilde{C}^{\top})^{-1}\hat{\beta}$$

$$= Y - Z\hat{\beta}$$

$$= \hat{\epsilon}_z$$

(3)

$$R_w^2 = 1 - \frac{\|\hat{\epsilon}_w\|^2}{\sum_{i=1}^n (Y_i - \bar{Y})^2}$$
$$= 1 - \frac{\|\hat{\epsilon}_z\|^2}{\sum_{i=1}^n (Y_i - \bar{Y})^2}$$
$$= R_z^2$$

(4)

$$\hat{\sigma}_w^2 = \frac{\|\hat{\epsilon}_w\|^2}{n - r - 1}$$
$$= \frac{\|\hat{\epsilon}_z\|^2}{n - r - 1}$$
$$= \hat{\sigma}_z^2$$

Method 1:

Let  $R = [\vec{0} \ I_r]$ , then it is equivalent to test  $H_0 : R\beta = \vec{0}$  and  $H_0 : R\gamma = \vec{0}$ .

$$R(\tilde{C}^{\top})^{-1}\hat{\beta} = \begin{bmatrix} \vec{0} & I_r \end{bmatrix} \begin{bmatrix} 1 & \vec{0}^{\top} \\ \vec{0} & (C^{-1})^{\top} \end{bmatrix} \begin{bmatrix} \hat{\beta}_0 \\ \hat{\beta}_{(2)} \end{bmatrix}$$
$$= (C^{-1})^{\top}\hat{\beta}_{(2)}$$

where 
$$\hat{\beta}_{(2)} = \begin{bmatrix} \hat{\beta}_1 \\ \hat{\beta}_2 \\ \vdots \\ \hat{\beta}_r \end{bmatrix}$$

And

$$(R((Z\tilde{C}^{\top})^{\top}(Z\tilde{C}^{\top}))^{-1}R^{\top})^{-1}$$

$$= (R(\tilde{C}Z^{\top}Z\tilde{C}^{\top})^{-1}R^{\top})^{-1}$$

$$= (R(\tilde{C}^{\top})^{-1}(Z^{\top}Z)^{-1}\tilde{C}^{-1}R^{T})^{-1}$$

$$= ([\vec{0} I_{r}] \begin{bmatrix} 1 & \vec{0}^{\top} \\ \vec{0} & (C^{-1})^{\top} \end{bmatrix} \begin{bmatrix} \Omega_{11} & \Omega_{12} \\ \Omega_{21} & \Omega_{22} \end{bmatrix} \begin{bmatrix} 1 & \vec{0}^{\top} \\ \vec{0} & C^{-1} \end{bmatrix} \begin{bmatrix} \vec{0}^{\top} \\ I_{r} \end{bmatrix})^{-1}$$

$$= ((C^{-1})^{\top}\Omega_{22}C^{-1})^{-1}$$

$$= C\Omega_{22}^{-1}C^{\top}$$

Therefore, the test statistic for  $H_0: R\gamma = \vec{0}$  is

$$\begin{split} &\frac{1}{\hat{\sigma}_{w}^{2}}(R\hat{\gamma})^{\top}(R(W^{\top}W)^{-1}R^{\top})^{-1}(R\hat{\gamma}) \\ &= \frac{1}{\hat{\sigma}_{z}^{2}}(R(\tilde{C}^{\top})^{-1}\hat{\beta})^{\top}(R((Z\tilde{C}^{\top})^{\top}(Z\tilde{C}^{\top}))^{-1}R^{\top})^{-1}(R(\tilde{C}^{\top})^{-1}\hat{\beta}) \\ &= \frac{1}{\hat{\sigma}_{z}^{2}}((C^{-1})^{\top}\hat{\beta}_{(2)})^{\top}C\Omega_{22}^{-1}C^{\top}((C^{-1})^{\top}\hat{\beta}_{(2)}) \\ &= \frac{1}{\hat{\sigma}_{z}^{2}}\hat{\beta}_{(2)}^{\top}\Omega_{22}^{-1}\hat{\beta}_{(2)} \end{split}$$

which is the same as the test statistic for  $H_0: R\beta = \vec{0}$ .

#### Method 2:

Here we use the residuals to derive the F-test statistic. We can easily find that

$$\hat{\epsilon}_{w,(red)} = \hat{\epsilon}_{z,(red)}$$

because the reduced models are the same.

Then the test statistic for  $H_0: \gamma_1 = \gamma_2 = ... = \gamma_r = 0$  is

$$\begin{split} &\frac{1}{\hat{\sigma}_w^2} (\|\hat{\epsilon}_{w,(red)}\|^2 - \|\hat{\epsilon}_w\|^2) \\ &= \frac{1}{\hat{\sigma}_z^2} (\|\hat{\epsilon}_{z,(red)}\|^2 - \|\hat{\epsilon}_z\|^2) \end{split}$$

which is the same as the test statistic for  $H_0: \beta_1 = \beta_2 = ... = \beta_r = 0$ .

(5)

$$\vec{w}_0 = \begin{bmatrix} 1 \\ w_{01} \\ \vdots \\ w_{0r} \end{bmatrix}$$
$$= \begin{bmatrix} 1 & \vec{0}^{\top} \\ \vec{0} & C \end{bmatrix} \begin{bmatrix} 1 \\ z_{01} \\ \vdots \\ z_{0r} \end{bmatrix}$$
$$= \tilde{C} \vec{z}_0$$

The center for the prediction interval for  $Y_0$  based on  $\hat{\gamma}$  is

$$\vec{w}_0^{\top} \hat{\gamma} = (\tilde{C} \vec{z}_0)^{\top} (\tilde{C}^{\top})^{-1} \hat{\beta}$$
$$= \vec{z}_0^{\top} \tilde{C}^{\top} (\tilde{C}^{\top})^{-1} \hat{\beta}$$
$$= \vec{z}_0^{\top} \hat{\beta}$$

which is the same as the center for the prediction interval for  $Y_0$  based on  $\hat{\beta}$ .

The half length of the prediction interval for  $Y_0$  based on  $\hat{\gamma}$  is

$$\begin{split} &\hat{\sigma}_w t_{n-r-1}(\frac{\alpha}{2})\sqrt{1+\vec{w}_0^\top (W^\top W)^{-1}\vec{w}_0} \\ &= \hat{\sigma}_z t_{n-r-1}(\frac{\alpha}{2})\sqrt{1+(\tilde{C}\vec{z}_0)^\top ((Z\tilde{C}^\top)^\top (Z\tilde{C}^\top))^{-1}(\tilde{C}\vec{z}_0)} \\ &= \hat{\sigma}_z t_{n-r-1}(\frac{\alpha}{2})\sqrt{1+\vec{z}_0^\top \tilde{C}^\top (\tilde{C}Z^\top Z\tilde{C}^\top)^{-1}\tilde{C}\vec{z}_0} \\ &= \hat{\sigma}_z t_{n-r-1}(\frac{\alpha}{2})\sqrt{1+\vec{z}_0^\top \tilde{C}^\top (\tilde{C}^\top)^{-1}(Z^\top Z)^{-1}\tilde{C}^{-1}\tilde{C}\vec{z}_0} \\ &= \hat{\sigma}_z t_{n-r-1}(\frac{\alpha}{2})\sqrt{1+\vec{z}_0^\top (Z^\top Z)^{-1}\vec{z}_0} \end{split}$$

which is the same as the half length for the prediction interval for  $Y_0$  based on  $\hat{\beta}$ . In all, the prediction interval for  $Y_0$  based on  $\hat{\beta}$  and the one based on  $\hat{\gamma}$  are the same.

#### Question 5

(1)

$$\hat{\beta} = \begin{bmatrix} 30.967 \\ 2.634 \\ 0.045 \end{bmatrix}$$

(2)

$$R^2 = 0.834$$

(3)

$$\hat{\sigma}^2 = 12.059$$

$$\begin{split} \hat{Cov}(\hat{\beta}) &= \hat{\sigma}^2 (Z^\top Z)^{-1} \\ &= \begin{bmatrix} 62.129 & 3.068 & -1.765 \\ 3.068 & 0.617 & -0.207 \\ -1.765 & -0.207 & 0.081 \end{bmatrix} \end{split}$$

(4) The 95% confidence interval for  $\beta_1$  is

(5) The 95% confidence intervals for  $\beta_j,\ j=0,1,2$  based on confidence region are

$$\beta_0 \in [6.557, 55.376]$$
  
 $\beta_1 \in [0.202, 5.067]$   
 $\beta_2 \in [-0.838, 0.928]$ 

(6) The 95% confidence intervals for  $\beta_j$ , j=0,1,2 based on Bonferroni correction are

$$\beta_0 \in [10.039, 51.894]$$
  
 $\beta_1 \in [0.549, 4.720]$   
 $\beta_2 \in [-0.712, 0.802]$ 

(7) The F-test statistic is

$$\frac{1}{\hat{\sigma}^2} (C\hat{\beta})^{\top} (C(Z^{\top}Z)^{-1}C^{\top})^{-1} (C\hat{\beta})$$
  
= 85.655

The critical value is

$$(r-q)F_{r-q,n-r-1}(\alpha) = (2-0)F_{2-0,20-2-1}(0.05)$$
  
= 7.183

As 85.655 > 7.183, we reject the  $H_0$ .

(8)

$$\bar{z}_1 = 16.222$$
 $\bar{z}_2 = 63.065$ 

So

$$\vec{z}_0 = \begin{bmatrix} 1\\16.222\\63.065 \end{bmatrix}$$

The confidence interval for the mean response  $E(Y_0)$  given  $\vec{z}_0$  is given by

$$[\vec{z}_0^{\top} \hat{\beta} - \hat{\sigma} t_{n-r-1} (\frac{\alpha}{2}) \sqrt{\vec{z}_0^{\top} (Z^{\top} Z)^{-1} \vec{z}_0}, \vec{z}_0^{\top} \hat{\beta} + \hat{\sigma} t_{n-r-1} (\frac{\alpha}{2}) \sqrt{\vec{z}_0^{\top} (Z^{\top} Z)^{-1} \vec{z}_0}]$$

The result is

(9) The prediction interval for the  $Y_0$  given  $\vec{z}_0$  is given by

$$[\vec{z}_0^{\top} \hat{\beta} - \hat{\sigma} t_{n-r-1} (\frac{\alpha}{2}) \sqrt{1 + \vec{z}_0^{\top} (Z^{\top} Z)^{-1} \vec{z}_0}, \vec{z}_0^{\top} \hat{\beta} + \hat{\sigma} t_{n-r-1} (\frac{\alpha}{2}) \sqrt{1 + \vec{z}_0^{\top} (Z^{\top} Z)^{-1} \vec{z}_0}]$$

The result is