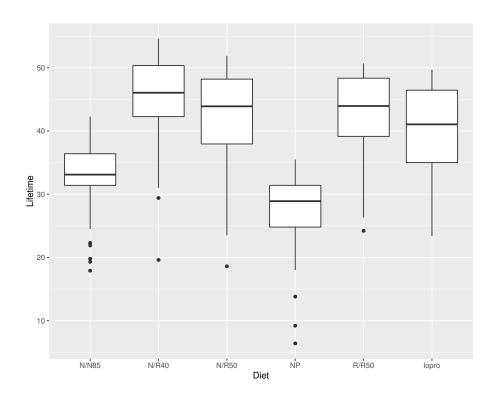
## STAT 510 Homework 3

## Yifan Zhu

## January 31, 2017

## **1.** (a)



$$SSE_{full} = 15297$$

(c) 
$$\hat{\sigma^2} = \frac{SSE_{full}}{n-r} = \frac{15297}{343} = 44.60$$

(d) 
$$SSE_{reduced} = 15511$$

(e) 
$$F = \frac{(SSE_{reduced} - SSE_{full})/(DFE_{reduced} - DFE_{full})}{SSE_{full}/DFE_{full}} = \frac{(15511 - 15297)/1}{15297/343} = 4.8$$

(f) This test is to test the among the mice having these six diet plans, wheatehr the mice having diet plan N/R50 and N/R50 lopro have the same population life time.

(g) If the parameter vector  $\boldsymbol{\beta} = \begin{bmatrix} \beta_1 & \beta_2 & \beta_2 & \beta_4 & \beta_5 & \beta_6 \end{bmatrix}$  are in the order for N/N85, N/R40, N/R50, NP, R/R50 and N/R50 lopro, then

$$C = \begin{bmatrix} 0 & 0 & 1 & 0 & 0 & -1 \end{bmatrix}, d = 0$$

Then

$$F = \frac{(\boldsymbol{C}\hat{\boldsymbol{\beta}} - d)^T (\boldsymbol{C}(\boldsymbol{X}^T \boldsymbol{X}) - \boldsymbol{C}^T)^{-1} (\boldsymbol{C}\hat{\boldsymbol{\beta}} - d) / rank(\boldsymbol{C})}{\hat{\sigma}^2} = 4.8$$

where X is the model matrix for full model.

The F statistic is the same as is computed in (e).

2. Let 
$$X = \begin{bmatrix} 1 & 1 \end{bmatrix}$$
, then  $A = X^T X = \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}$  is symmetric. Let  $G = \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}$ , then 
$$AGA = \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} = A$$

Thus G is a generalized inverse of the symmetric metrix A, but it is not symmetric.

**3.** (a)

$$\begin{aligned} rank(\boldsymbol{X}^T\boldsymbol{X}) &\leq rank(\boldsymbol{X}) \\ rank(\boldsymbol{X}) &= rank(\boldsymbol{P}_{\boldsymbol{X}}) = rank(\boldsymbol{X}(\boldsymbol{X}^T\boldsymbol{X})^-\boldsymbol{X}^T\boldsymbol{X}) \leq rank(\boldsymbol{X}^T\boldsymbol{X}) \\ \Rightarrow rank(\boldsymbol{X}) &= rank(\boldsymbol{X}^T\boldsymbol{X}) \end{aligned}$$

(b)

$$rank(\mathbf{X}) = rank(\mathbf{P}_{\mathbf{X}}\mathbf{X}) \le rank(\mathbf{P}_{\mathbf{X}})$$
$$rank(\mathbf{P}_{\mathbf{X}}) = rank(\mathbf{X}(\mathbf{X}^{T}\mathbf{X})^{-}\mathbf{X}) \le rank(\mathbf{X})$$
$$\Rightarrow rank(\mathbf{X}) = rank(\mathbf{P}_{\mathbf{X}})$$

(c) First we can prove  $rank(\mathbf{AP_X}) = rank(\mathbf{AP_X})$ .

$$rank(\mathbf{A}\mathbf{X}) = rank(\mathbf{A}\mathbf{P}_{\mathbf{X}}\mathbf{X}) \leq rank(\mathbf{A}\mathbf{P}_{\mathbf{X}})$$
$$rank(\mathbf{A}\mathbf{P}_{\mathbf{X}}) = rank(\mathbf{A}\mathbf{X}(\mathbf{X}^{T}\mathbf{X})^{-}\mathbf{X}^{T}) \leq rank(\mathbf{A}\mathbf{X})$$
$$\Rightarrow rank(\mathbf{A}\mathbf{X}) = rank(\mathbf{A}\mathbf{P}_{\mathbf{X}})$$

Then we have

$$\boldsymbol{C}(\boldsymbol{X}^T\boldsymbol{X})^-\boldsymbol{C}^T = \boldsymbol{A}\boldsymbol{X}(\boldsymbol{X}^T\boldsymbol{X})^-\boldsymbol{X}^T\boldsymbol{A}^T = \boldsymbol{A}\boldsymbol{P}_{\boldsymbol{X}}\boldsymbol{A}^T = \boldsymbol{A}\boldsymbol{P}_{\boldsymbol{X}}\boldsymbol{P}_{\boldsymbol{X}}\boldsymbol{A}^T$$
 Thus  $rank(\boldsymbol{C}(\boldsymbol{X}^T\boldsymbol{X})^-\boldsymbol{C}^T) = rank(\boldsymbol{A}\boldsymbol{P}_{\boldsymbol{X}}(\boldsymbol{A}\boldsymbol{P}_{\boldsymbol{X}})^T) = rank(\boldsymbol{A}\boldsymbol{P}_{\boldsymbol{X}}) = rank(\boldsymbol{A}\boldsymbol{X}) = rank(\boldsymbol{C}) = q.$ 

- (d) From (c) we know  $rank(\mathbf{A}\mathbf{X}) = rank(\mathbf{A}\mathbf{P}_{\mathbf{X}}) = rank(\mathbf{A})$ .
- 4. (a) Let the parameter vector in such an order

$$\begin{array}{l} (S1,F1,25\%) \\ (S2,F1,25\%) \\ (S1,F2,25\%) \\ (S2,F2,25\%) \\ (S1,F1,50\%) \\ (S2,F1,50\%) \\ (S1,F2,50\%) \\ (S1,F2,50\%) \\ (S2,F2,50\%) \\ (S1,F1,75\%) \\ (S2,F1,75\%) \\ (S1,F2,75\%) \\ (S2,F2,75\%) \\ (S2,F2,75\%) \\ \end{array}$$

Then

$$se_{(a)} = 11.59$$

(b) 
$$C_{(b)} = \begin{bmatrix} 1/6 & 1/6 & 0 & 0 & 1/6 & 1/6 & 0 & 0 & 1/6 & 1/6 & 0 & 0 \\ 0 & 0 & 1/6 & 1/6 & 0 & 0 & 1/6 & 1/6 & 0 & 0 & 1/6 & 1/6 \end{bmatrix}, \begin{bmatrix} \hat{\mu}_{F1,...} \\ \hat{\mu}_{F2,...} \end{bmatrix} = C_{(b)} \hat{\beta} = \begin{bmatrix} 214.75 \\ 181.08 \end{bmatrix}$$

(c) 
$$C_{(c)} = \begin{bmatrix} 0 & 1/2 & 0 & 1/2 & 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix}, \hat{\mu}_{.,S2,25\%} = C_{(c)}\hat{\beta} = 156.25$$

(d) 
$$se_{(c)} = \sqrt{\hat{\sigma}^2 C_{(c)}(X^T X)^- C_{(c)}^T} = 8.20$$

(e) 
$$H_0: \bar{\mu}_{F1,...} = \bar{\mu}_{F2,...}(\mathbf{C}\boldsymbol{\beta} - d = 0)$$

$$H_a: \bar{\mu}_{F1,...} \neq \bar{\mu}_{F2,...}(\mathbf{C}\boldsymbol{\beta} - d \neq 0)$$

$$C = \begin{bmatrix} 1/6 & 1/6 & -1/6 & -1/6 & 1/6 & 1/6 & -1/6 & -1/6 & 1/6 & -1/6 & -1/6 \end{bmatrix}, d = 0$$

$$F = \frac{(C\hat{\beta} - d)^T (C(X^T X)^- C^T)^{-1} (C\hat{\beta} - d) / rank(C)}{\hat{\sigma}^2} = 25.3$$

p-value = 0.0003

Conclusion: Since p-value is small, we decide to reject the null hypothesis and conclude that there is filler type main effect for the fabric loss.

(f)

 $H_0$ : there is no three way interaction.  $(C\beta - d = 0)$  $H_a$ : there is three way interaction.  $(C\beta - d \neq 0)$ 

$$\begin{split} \pmb{C} &= \begin{bmatrix} 1 & -1 & -1 & 1 & -1 & 1 & 1 & -1 & 0 & 0 & 0 & 0 \\ 1 & -1 & -1 & 1 & 0 & 0 & 0 & 0 & -1 & 1 & 1 & -1 \end{bmatrix}, \, \pmb{d} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \\ F &= \frac{(\pmb{C}\hat{\pmb{\beta}} - d)^T (\pmb{C}(\pmb{X}^T \pmb{X})^- \pmb{C}^T)^{-1} (\pmb{C}\hat{\pmb{\beta}} - d)/rank(\pmb{C})}{\hat{\sigma}^2} = 0.89 \\ \text{p-value} &= 0.44 \end{split}$$

Conclusion: Since p-value is big, we fail to reject the null hypothesis and conclude that there is no three way interaction among the three factors for fabric loss.

(g)

 $H_0$ : there is no two way interaction between filler type and filler proportion.  $(C\beta - d = 0)$  $H_a$ : there is two way interaction between filler type and filler proportion.  $(C\beta - d \neq 0)$ 

$$\begin{aligned} & \boldsymbol{C} = \begin{bmatrix} 1/2 & 1/2 & -1/2 & -1/2 & -1/2 & -1/2 & 1/2 & 1/2 & 0 & 0 & 0 & 0 \\ 1/2 & 1/2 & -1/2 & -1/2 & 0 & 0 & 0 & 0 & -1/2 & -1/2 & 1/2 & 1/2 \end{bmatrix}, \, \boldsymbol{d} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \\ F & = \frac{(\boldsymbol{C}\hat{\boldsymbol{\beta}} - d)^T (\boldsymbol{C}(\boldsymbol{X}^T\boldsymbol{X})^- \boldsymbol{C}^T)^{-1} (\boldsymbol{C}\hat{\boldsymbol{\beta}} - d)/rank(\boldsymbol{C})}{\hat{\sigma}^2} = 6.57 \end{aligned}$$

p-value = 0.01

Conclusion: Since p-value is big, we decide to reject the null hypothesis and conclude that there is two way interaction between filler type and filler proportion for fabric loss.