## Homework Key #4

**1** See the key of the midterm.

**2**. See the key of the sample midterm.

**3** Consider the following model:

$$Y_1 = \tau_1 + \tau_2 + \tau_3 + \epsilon_1$$
  

$$Y_2 = \tau_1 + \tau_3 + \epsilon_2$$
  

$$Y_3 = \tau_2 + \epsilon_3$$

(a) Write out the model in matrix form. What is the rank of the design matrix?

(b) Which (if any) of the individual parameters are estimable?

(c) Is  $\tau_1 - 2\tau_2 + \tau_3$  estimable? Explain how you know.

(d) Find the BLUE of  $\tau_1 - 2\tau_2 + \tau_3$ . In addition, find another estimate of  $\tau_1 - 2\tau_2 + \tau_3$  that is not the BLUE.

Solution: (a) We can write

$$\begin{bmatrix} Y_1 \\ Y_2 \\ Y_3 \end{bmatrix} = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} \tau_1 \\ \tau_2 \\ \tau_3 \end{bmatrix} + \begin{bmatrix} \epsilon_1 \\ \epsilon_2 \\ \epsilon_3 \end{bmatrix}.$$
  
Letting  $\mathbf{Y} = \begin{bmatrix} Y_1 \\ Y_2 \\ Y_3 \end{bmatrix}$ ,  $\mathbf{X} = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix}$ ,  $\beta = \begin{bmatrix} \tau_1 \\ \tau_2 \\ \tau_3 \end{bmatrix}$ ,  $\epsilon = \begin{bmatrix} \epsilon_1 \\ \epsilon_2 \\ \epsilon_3 \end{bmatrix}$ , we can also write  $\mathbf{Y} = \mathbf{X}\beta + \epsilon$ .

The Gaussian elimination gives the rank of the design matrix  $\mathbf{X}$  of 2.

(b) Let  $\mathbf{e}_i \in \mathbb{R}^3$  be the vector the *i*th element of which is 1 and the rest of which is 0. Then  $\mathbf{e}_i^T \beta = \tau_i$ . Define the augmented matrices given by  $[\mathbf{X}^T \mathbf{e}_i]$ . The Gaussian elimination shows that  $\mathbf{e}_1 \notin \mathcal{R}(\mathbf{X}^T)$ ,  $\mathbf{e}_3 \notin \mathcal{R}(\mathbf{X}^T)$  but  $\mathbf{e}_2 \in \mathcal{R}(\mathbf{X}^T)$ . Applying the Lemma 10.2.1, we conclude that  $\tau_2$  is estimable but that other two parameters are not.

(c) Let  $\mathbf{c} = (1, -2, 1)^T$ . Then  $\mathbf{c}^T \boldsymbol{\beta} = \tau_1 - 2\tau_2 + \tau_3$ . As in the previous problem, define the augmented matrix  $[\mathbf{X}^T \mathbf{c}]$  and perform the Gaussian elimination to conclude that  $\tau_1 - 2\tau_2 + \tau_3$  is estimable by the Lemma 10.2.1.

(d) Let

$$\mathbf{X}_1 = \begin{bmatrix} 1 & 1\\ 1 & 0\\ 0 & 1 \end{bmatrix} \text{ and } \mathbf{X}_2 = \begin{bmatrix} 1\\ 1\\ 0 \end{bmatrix}.$$

Note that  $\mathbf{X} = [\mathbf{X}_1 \ \mathbf{X}_2]$ . Note also that  $\mathbf{X}$  has two independent columns and those correspond to columns of  $\mathbf{X}_1$ . Thus, a generalized inverse of  $\mathbf{X}^T \mathbf{X}$  is given by

$$\mathbf{X}^T \mathbf{X}^- = \left[egin{array}{ccc} (\mathbf{X}_1^T \mathbf{X}_1)^{-1} & \mathbf{0} \ \mathbf{0} & \mathbf{0} \end{array}
ight]$$

with

$$(\mathbf{X}_1^T \mathbf{X}_1)^{-1} = \frac{1}{3} \begin{bmatrix} 2 & -1 \\ -1 & 2 \end{bmatrix}$$

This gives a least squares estimate  $\hat{\beta}$  of  $\beta$  by

$$\hat{\beta} \equiv (\mathbf{X}^T \mathbf{X})^{-} \mathbf{X}^T \mathbf{Y} = \frac{1}{3} \begin{bmatrix} Y_1 + 2Y_2 - Y_3 \\ Y_1 - Y_2 + 2Y_3 \\ 0 \end{bmatrix}.$$

Because  $\mathbf{c}^T \boldsymbol{\beta}$  is estimable by part(c), the estimate  $\mathbf{c}^T \hat{\boldsymbol{\beta}}$  given by

$$\mathbf{c}^T \hat{\beta} = \frac{-Y_1 + 4Y_2 - 5Y_3}{3}$$

is the BLUE of  $\tau_1 - 2\tau_2 + \tau_3$  by the Gauss-Markov Theorem. A possible linear unbiased estimate of  $\tau_1 - 2\tau_2 + \tau_3$  is  $Y_2 - 2Y_3$  because

$$E[Y_2 - 2Y_3] = E[Y_2] - 2E[Y_3] = (\tau_1 + \tau_3) - 2\tau_2 = \tau_1 - 2\tau_2 + \tau_3$$

and

$$Y_2 - 2Y_3 = \mathbf{d}^T \mathbf{Y}$$

where  $\mathbf{d} = (0, 1, -2)^T$ . This is not the BLUE of  $\tau_1 - 2\tau_2 + \tau_3$  because the BLUE is unique by the Gauss-Markov Theorem.