

2. a) Show that  $\begin{pmatrix} \beta_1 + \beta_2 \\ \beta_1 - 2\beta_2 \end{pmatrix}$  is estimable.

To show estimable, either:

1) Show  $X$  full rank

OR

2) Show  $\Lambda' = P'X$

1) If we row reduce  $X$ , get  $\begin{pmatrix} 1 & 0 \\ 0 & 1 \\ 0 & 0 \\ 0 & 0 \end{pmatrix} \Rightarrow \text{rank}(X) = \# \text{ pivots} = 2$

Know  $X$  full rank if  $\text{rank}(X) = \min(\# \text{ rows}, \# \text{ cols})$ .

Since  $\text{rank}(X) = 2 = \min(\underbrace{\# \text{ rows}}_4, \underbrace{\# \text{ cols}}_2) = 2$ , then  $X$  full rank  $\Rightarrow$

any  $\Lambda'\beta$  is estimable  $\Rightarrow \begin{pmatrix} \beta_1 + \beta_2 \\ \beta_1 - 2\beta_2 \end{pmatrix}$  estimable.

2) However, this approach is the way to go b/c we will need a projection matrix for part b).

$$\Lambda = \begin{pmatrix} 1 & 1 \\ 1 & -2 \end{pmatrix} \quad \text{and} \quad X = \begin{pmatrix} 1 & 3 \\ 1 & 1 \\ 1 & 1 \\ 1 & 2 \end{pmatrix}. \quad \text{Then} \quad \begin{pmatrix} 1 & 1 \\ 1 & -2 \end{pmatrix} = \begin{pmatrix} p_1 & p_2 & p_3 & p_4 \\ p_5 & p_6 & p_7 & p_8 \end{pmatrix} \begin{pmatrix} 1 & 3 \\ 1 & 1 \\ 1 & 1 \\ 1 & 2 \end{pmatrix}$$

$$\Rightarrow 1 = p_1 + p_2 + p_3 + p_4 \quad 1 = 3p_1 + p_2 + p_3 + 2p_4$$

$$1 = p_5 + p_6 + p_7 + p_8 \quad -2 = 3p_5 + p_6 + p_7 + 2p_8 \Rightarrow \begin{array}{l} 1 = p_5 + p_6 + p_7 + p_8 \\ -2 = 3p_5 + p_6 + p_7 + 2p_8 \end{array} \Rightarrow \begin{array}{l} 1 = p_5 + p_6 + p_7 + p_8 \\ -2 = 3p_5 + p_6 + p_7 + 2p_8 \end{array}$$

$$\text{Let } p_1 = p_2 = p_4 = 0 \Rightarrow p_3 = 1$$

$$\text{Let } p_5 = -4 \Rightarrow p_8 = 5 \quad \text{and let } p_6 = p_7 = 0$$

$$\text{Thus, } \Lambda' = P'X \quad \text{for } P = \begin{pmatrix} 0 & -4 \\ 0 & 0 \\ 1 & 0 \\ 0 & 5 \end{pmatrix} \Rightarrow \Lambda'\beta \text{ estimable}$$

$$\text{Check } \begin{pmatrix} 1 & 1 \\ 1 & -2 \end{pmatrix} \stackrel{?}{=} \begin{pmatrix} 0 & 0 & 1 & 0 \\ -4 & 0 & 0 & 5 \end{pmatrix} \begin{pmatrix} 1 & 3 \\ 1 & 1 \\ 1 & 1 \\ 1 & 2 \end{pmatrix} = \begin{pmatrix} 1 & 1 \\ 1 & -2 \end{pmatrix} \checkmark$$

2 b) Find the UMVUE of  $\begin{pmatrix} \beta_1 + \beta_2 \\ \beta_1 - 2\beta_2 \end{pmatrix}$

By Gauss Markov, we know the BLUE (UMVUE) of  $\Lambda'\beta$  is  $P'MY$ .

From part a), know one projection matrix is  $P = \begin{pmatrix} 0 & -4 \\ 0 & 0 \\ 0 & 0 \\ 0 & 5 \end{pmatrix}$ .

Now, need to find  $M$ .

$$(X'X)^{-1} = \left[ \begin{pmatrix} 1 & 1 & 1 & 1 \\ 3 & 1 & 1 & 2 \end{pmatrix} \begin{pmatrix} 1 & 3 \\ 1 & 1 \\ 1 & 1 \\ 2 & 1 \end{pmatrix} \right]^{-1} = \begin{bmatrix} 4 & 7 \\ 7 & 15 \end{bmatrix}^{-1} = \frac{1}{11} \begin{bmatrix} 15 & -7 \\ -7 & 4 \end{bmatrix}$$

$$\text{Then, } M = X(X'X)^{-1}X' = \frac{1}{11} \begin{pmatrix} 13 \\ 11 \\ 11 \\ 12 \end{pmatrix} \begin{bmatrix} 15 & -7 \\ -7 & 4 \end{bmatrix} \begin{pmatrix} 1 & 1 & 1 & 1 \\ 3 & 1 & 1 & 2 \end{pmatrix} = \frac{1}{11} \begin{pmatrix} -6 & 5 \\ 8 & -3 \\ 8 & -3 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} 1 & 1 & 1 & 1 \\ 3 & 1 & 1 & 2 \end{pmatrix}$$

UMVUE

$$= \frac{1}{11} \begin{pmatrix} 9 & -1 & -1 & 4 \\ -1 & 5 & 5 & 2 \\ -1 & 5 & 5 & 2 \\ 4 & 2 & 2 & 3 \end{pmatrix}. \text{ Then, } \Lambda'\hat{\beta} = P'MY = \frac{1}{11} \begin{pmatrix} 0 & 0 & 1 & 0 \\ -4 & 0 & 0 & 5 \end{pmatrix} \begin{pmatrix} 9 & -1 & -1 & 4 \\ -1 & 5 & 5 & 2 \\ -1 & 5 & 5 & 2 \\ 4 & 2 & 2 & 3 \end{pmatrix} Y$$

$$= \frac{1}{11} \begin{pmatrix} -1 & 5 & 5 & 2 \\ -16 & 14 & 14 & -1 \end{pmatrix} \begin{pmatrix} y_1 \\ y_2 \\ y_3 \\ y_4 \end{pmatrix}$$

$$= \frac{1}{11} \begin{pmatrix} -y_1 + 5y_2 + 5y_3 + 2y_4 \\ -16y_1 + 14y_2 + 14y_3 - y_4 \end{pmatrix}$$

## 2014 Section 2 Quid

2c) Find the distr. of the UMVUE in b)

$$\begin{aligned}
 \Gamma \quad \text{Know } \underbrace{\Lambda' \hat{\beta}}_{\text{UMVUE}} = P'MY &\sim N(E[P'MY], \text{Cov}(P'MY)) \\
 &\equiv N(\Lambda'\beta, P'M \text{Cov}(Y) (P'M)') \\
 &\equiv N(\Lambda'\beta, P'M \sigma^2 I_M P) \\
 &\equiv N(\Lambda'\beta, \sigma^2 P'MP)
 \end{aligned}$$

Know from b)  $P'M = \frac{1}{11} \begin{pmatrix} -1 & 5 & 5 & 2 \\ -16 & 14 & 14 & -1 \end{pmatrix}$

$$\begin{aligned}
 \text{Since } P'MP &= (P'M)(P'M)' = \frac{1}{121} \begin{pmatrix} -1 & 5 & 5 & 2 \\ -16 & 14 & 14 & -1 \end{pmatrix} \begin{pmatrix} -1 & -16 \\ 5 & 14 \\ 5 & 14 \\ 2 & -1 \end{pmatrix} \\
 &= \frac{1}{11} \begin{pmatrix} 5 & 14 \\ 14 & 59 \end{pmatrix}
 \end{aligned}$$

$$\Rightarrow P'MY \sim N\left(\begin{pmatrix} \beta_1 + \beta_2 \\ \beta_1 - 2\beta_1 \end{pmatrix}, \frac{\sigma^2}{11} \begin{pmatrix} 5 & 14 \\ 14 & 59 \end{pmatrix}\right)$$

Note Could have also recognized that:

$$\Lambda' \hat{\beta} = P'MY = N(\Lambda'\beta, \sigma^2 P'MP)$$

$$\text{Since } \Lambda' = P'X \Rightarrow P'MP = \underbrace{P'X}_{\Lambda'} (X'X)^{-1} \underbrace{X'P}_{\Lambda} = \Lambda' (X'X)^{-1} \Lambda$$

$$\Rightarrow P'MY = N(\Lambda'\beta, \sigma^2 \Lambda' (X'X)^{-1} \Lambda)$$

Know from b) that  $(X'X)^{-1} = \frac{1}{11} \begin{pmatrix} 15 & -7 \\ -7 & 4 \end{pmatrix}$  and  $\Lambda = \begin{pmatrix} 1 & 1 \\ 1 & -2 \end{pmatrix}$

$$\begin{aligned}
 \Rightarrow \Lambda' (X'X)^{-1} \Lambda &= \frac{1}{11} \begin{pmatrix} 1 & 1 \\ 1 & -2 \end{pmatrix} \begin{pmatrix} 15 & -7 \\ -7 & 4 \end{pmatrix} \begin{pmatrix} 1 & 1 \\ 1 & -2 \end{pmatrix} = \frac{1}{11} \begin{pmatrix} 8 & -3 \\ 29 & -15 \end{pmatrix} \begin{pmatrix} 1 & 1 \\ 1 & -2 \end{pmatrix} \\
 &= \frac{1}{11} \begin{pmatrix} 5 & 14 \\ 14 & 59 \end{pmatrix}
 \end{aligned}$$

29  
50  
59

2.1) Now  $E(\epsilon) = 0$  and  $\text{Cov}(\epsilon) = \Sigma = \sigma^2 V \sim I \sim V \text{ P.D.}$

I identify a model, call it model (2), such that model (2) is in the form of an ordinary linear model (i.e., for the error term, the mean is 0 and variance is  $\sigma^2 I$ ) and it contains the same parameters as in model (1).

Have Model (1):  $y = X\beta + \epsilon$  where  $\epsilon \sim N(0, \sigma^2 V)$

Let  $V = QQ'$  where  $Q$  invertible since  $V \text{ P.D.}$

Then,  $\underbrace{Q^{-1}y}_{y^*} = \underbrace{Q^{-1}X\beta}_{X^*} + \underbrace{Q^{-1}\epsilon}_{\epsilon^*} \Rightarrow y^* = X^*\beta + \epsilon^*$  where  $E[\epsilon^*] = E[Q^{-1}\epsilon] = Q^{-1}E[\epsilon] = 0$

$$\begin{aligned} \text{Cov}[\epsilon^*] &= \text{Cov}[Q^{-1}\epsilon] = Q^{-1}\text{Cov}(\epsilon)Q^{-1} \\ &= Q^{-1}\sigma^2 V Q^{-1} \\ &= \sigma^2 Q^{-1}(QQ')Q^{-1} \\ &= \sigma^2 I \cdot I = \sigma^2 I \end{aligned}$$

2.e) Show that  $\lambda'\beta$  is estimable in model (1) iff  $\lambda'\beta$  is estimable in model (2).

( $\Rightarrow$ ) Assume  $\lambda'\beta$  estimable in model (1)  $\Rightarrow \lambda' = P'X \Rightarrow \lambda' = P' \underbrace{Q^{-1}Q}_{I} X$

$$\Rightarrow \lambda' = \underbrace{P'Q}_{P^*}' \underbrace{Q^{-1}X}_{X^*} \Rightarrow \lambda' = P^* \underbrace{Q^{-1}X}_{X^*} \Rightarrow \lambda' = P^* X^* \Rightarrow \lambda'\beta \text{ estimable in model (2).}$$

( $\Leftarrow$ ) Assume  $\lambda'\beta$  estimable in model (2)  $\Rightarrow \lambda' = P^* X^* \Rightarrow \lambda' = \underbrace{P^* Q^{-1}}_{P^*}' X \Rightarrow \lambda' = P^* X$

$\Rightarrow \lambda'\beta$  estimable in model (1).  $\square$

2014, Section 2, Qund

2f. Show that  $Y^T \Sigma^{-1} Y - Y_1^2 / \sigma_{11}^2 \sim \chi_3^2$  given that  $\beta = 0$  and  $\Sigma = (\sigma_{ij})$ .

$$Y^T \Sigma^{-1} Y = \frac{1}{\sigma^2} Y' (Q Q')^{-1} Y = \frac{1}{\sigma^2} Y' Q'^{-1} Q^{-1} Y = \frac{1}{\sigma^2} \underbrace{(Q^{-1} Y)'}_{Y^*} \underbrace{(Q^{-1} Y)}_{Y^*}$$

$$\text{Since } \underbrace{Q^{-1} Y}_{Y^*} \sim N(0, \sigma^2 I) \Rightarrow \frac{1}{\sigma} Y^* \sim N(0, I) \\ \Rightarrow \frac{1}{\sigma^2} Y^{*'} Y^* \sim \chi_4^2$$

$$\text{Also, let } \Sigma_{11}^{-1} = \frac{1}{\sigma_{11}^2} \begin{bmatrix} 1 & 0 & \dots & 0 \\ 0 & 0 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 0 \end{bmatrix}$$

$$\text{Then, } \frac{Y_1^2}{\sigma_{11}^2} = Y' \Sigma_{11} Y. \text{ Since } Y^* = Q^{-1} Y \Rightarrow Q Y^* = Y$$

$$\Rightarrow Y' \Sigma_{11}^{-1} Y = (Q Y^*)' \Sigma_{11}^{-1} (Q Y^*) = Y^{*'} Q' \Sigma_{11}^{-1} Q Y^*$$

$$\text{Then, } \underbrace{Y' \Sigma^{-1} Y}_{\frac{1}{\sigma^2} Y^{*'} Y^*} - \underbrace{Y_1^2 / \sigma_{11}^2}_{Y^{*'} Q' \Sigma_{11}^{-1} Q Y^*} = \frac{1}{\sigma^2} Y^{*'} Y^* - Y^{*'} Q' \Sigma_{11}^{-1} Q Y^* \\ = Y^{*'} \left( \frac{1}{\sigma^2} I - Q' \Sigma_{11}^{-1} Q \right) Y^*$$

$$\sim \chi_3^2 \leftarrow \text{rank} \left( \frac{1}{\sigma^2} I - Q' \Sigma_{11}^{-1} Q \right) \quad \square$$