

## Homework 6 - Answers.

### 1. Question 5

$$(i) (AB)(AB)^{-1} = I \Rightarrow A^{-1}AB(AB)^{-1} = A^{-1} \Rightarrow B(AB)^{-1} = A^{-1} \Rightarrow \\ B^{-1}B(AB)^{-1} = B^{-1}A^{-1} \Rightarrow (AB)^{-1} = B^{-1}A^{-1}$$

$$(ii) (ABC)^{-1} = (AD)^{-1} \text{ if } D=BC. \text{ By part (i), } (AD)^{-1} = D^{-1}A^{-1}.$$

But again, by part (i),  $D^{-1} = (BC)^{-1} = C^{-1}B^{-1}$ . Hence,

$$(ABC)^{-1} = C^{-1}B^{-1}A^{-1}$$

### Question 8.

Property 5. (p.756) says that for  $A_{m \times n}$  and  $b_{n \times 1}$ , nonrandom,

$$V(Ay+b) = AV(y)A^T.$$

$$\text{By definition } V(Ay+b) = E\left\{[(Ay+b) - E(Ay+b)][(Ay+b) - E(Ay+b)]^T\right\} \\ = E\left\{[Ay+b - AE(y) - b][Ay+b - AE(y) - b]^T\right\} \\ = E\left\{A(y - E(y))[A(y - E(y))]^T\right\} \\ = E\left\{A(y - E(y))(y - E(y))^T A^T\right\} \\ = A E\left\{(y - E(y))(y - E(y))^T\right\} A^T \\ = AV(y)A^T.$$

2.  $AB = \begin{bmatrix} -35 & 16 \\ -56 & 28 \\ -73 & 38 \\ -48 & 36 \end{bmatrix}$

BA is not defined as the number of columns in B, i.e., 2 is different from the number of rows in A, i.e., 4

3. X is an  $n \times k$  matrix.

$$X^T = \begin{bmatrix} 1 & 1 & \dots & 1 \\ x_{12} & x_{22} & \dots & x_{n2} \\ \vdots & \vdots & & \vdots \\ x_{1k} & x_{2k} & \dots & x_{nk} \end{bmatrix}$$

$$X^T X = \begin{bmatrix} n & \sum_{i=1}^n x_{i2} & \sum_{i=1}^n x_{i3} & \dots & \sum_{i=1}^n x_{ik} \\ \sum_{i=1}^n x_{i2} & \sum_{i=1}^n x_{i2}^2 & \sum_{i=1}^n x_{i2}x_{i3} & \dots & \sum_{i=1}^n x_{i2}x_{ik} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ \sum_{i=1}^n x_{ik} & \sum_{i=1}^n x_{ik}x_{i2} & \sum_{i=1}^n x_{ik}x_{i3} & \dots & \sum_{i=1}^n x_{ik}^2 \end{bmatrix}$$

4. (a) let  $Y^T = (Y_1 \dots Y_n)_{1 \times n}$ ,  $X = \begin{bmatrix} 1 & X_{11} & X_{12} & X_{13} \\ 1 & X_{21} & X_{22} & X_{23} \\ \vdots & \vdots & \vdots & \vdots \\ 1 & X_{n1} & X_{n2} & X_{n3} \end{bmatrix}_{n \times 4}$ ,

$\beta^T = (\beta_0 \ \beta_1 \ \beta_2 \ \beta_3)_{1 \times 4}$  and  $U^T = (U_1 \dots U_n)$ . Then,

$$Y = X\beta + U.$$

(b)  $S_n(\beta) = Y^T Y - Y^T X \beta - \beta^T X^T Y + \beta^T X^T X \beta$   
 $= Y^T Y - 2Y^T X \beta + \beta^T X^T X \beta$

$$\frac{\partial S_n(\beta)}{\partial \beta} = -2Y^T X + 2\beta^T X^T X$$

(c)  $\frac{\partial S_n(\hat{\beta})}{\partial \beta} = 0 \Leftrightarrow -2Y^T X + 2\hat{\beta}^T X^T X = 0 \Leftrightarrow \hat{\beta}^T X^T X = Y^T X$   
 $\Leftrightarrow X^T X \hat{\beta} = X^T Y$

If  $X^T X$  is non-singular, i.e., if  $(X^T X)^{-1}$  exists,

$$\hat{\beta} = (X^T X)^{-1} X^T Y.$$

(d) Since,  $E(U) = \begin{bmatrix} E(U_1) \\ \vdots \\ E(U_n) \end{bmatrix}$ ,  $E(U) = 0$ , given that  $E(U_i) = 0$ .

$$E(UU^T) = E \left\{ \begin{bmatrix} U_1 \\ \vdots \\ U_n \end{bmatrix} [U_1 \dots U_n] \right\} = E \left\{ \begin{bmatrix} U_1^2 & U_1 U_2 & \dots & U_1 U_n \\ U_2 U_1 & U_2^2 & \dots & U_2 U_n \\ \vdots & \vdots & \ddots & \vdots \\ U_n U_1 & U_n U_2 & \dots & U_n^2 \end{bmatrix} \right\} = \sigma^2 I_n$$

Since  $E(U_i^2) = \sigma^2$  and  $\{U_i\}_{i=1}^n$  is an independent sequence.