

STAT5030 Assignment 4 Solution

1. (a) $\sum_{i=1}^n \hat{y}_i(y_i - \hat{y}_i) = HY(I - H)Y = 0.$
- (b) $\sum_{i=1}^n \text{Var}(\hat{y}_i) = \text{tr}(\text{Var}(HY)) = p\sigma^2.$

2. Note that

$$\begin{aligned}
 & M(\hat{\beta}, \beta) - M(\hat{\beta}_1, \beta) \\
 &= E(\hat{\beta} - \beta)(\hat{\beta} - \beta)^\top - E(\hat{\beta}_1 - \beta)(\hat{\beta}_1 - \beta)^\top \\
 &= (X^\top X)^{-1}\sigma^2 - \left(\frac{1}{\rho+1}\right)^2 (X^\top X)^{-1}\sigma^2 - \left(\frac{\rho}{\rho+1}\right)^2 \beta\beta^\top \\
 &= \frac{\rho^2 + 2\rho}{(\rho+1)^2} (X^\top X)^{-1}\sigma^2 - \frac{\rho^2}{(\rho+1)^2} \beta\beta^\top.
 \end{aligned}$$

Since \mathbf{X} has full column rank, for any vector \mathbf{v} , there exist a vector \mathbf{u} such that $\mathbf{v} = \rho^{-1}(1 + \rho)(\mathbf{X}^\top \mathbf{X})^{1/2}\mathbf{u}$, and

$$\begin{aligned}
 \mathbf{v}^\top (M(\hat{\beta}, \beta) - M(\hat{\beta}_1, \beta))\mathbf{v} &= \frac{(1+\rho)^2}{\rho^2} \mathbf{u}^\top (\mathbf{X}^\top \mathbf{X})^{1/2} (M(\hat{\beta}, \beta) - M(\hat{\beta}_1, \beta)) (\mathbf{X}^\top \mathbf{X})^{1/2} \mathbf{u} \\
 &\geq \sigma^2 \mathbf{I} - (\mathbf{X}^\top \mathbf{X})^{1/2} \beta \beta^\top (\mathbf{X}^\top \mathbf{X})^{1/2}.
 \end{aligned}$$

By the given theorem, $\sigma^2 \mathbf{I} - (\mathbf{X}^\top \mathbf{X})^{1/2} \beta \beta^\top (\mathbf{X}^\top \mathbf{X})^{1/2} \geq 0$ if and only if $\beta^\top \mathbf{X}^\top \mathbf{X} \beta \leq \sigma^2$.

3. (a) $E(\lambda^\top \hat{\beta}) = E(r^\top X^\top X \hat{\beta}) = E(r^\top X^\top X (X^\top X)^{-1} X^\top Y) = E(r^\top X^\top Y) = \lambda^\top \beta.$
- (b) $r^\top X^\top Y = \lambda^\top (X^\top X)^{-1} X^\top Y$. Since $\lambda^\top \beta$ is estimable, there exists a vector t such that $\lambda^\top \beta = t^\top X$. Then $r^\top X^\top Y = t^\top X (X^\top X)^{-1} X^\top Y$ is invariant to the choice of \mathbf{r} .

$$4. Y = X\beta + \varepsilon, \text{ where } X = \begin{pmatrix} 1 & 1 & 0 & 0 & 1 & 0 & 0 \\ 1 & 1 & 0 & 0 & 0 & 1 & 0 \\ 1 & 1 & 0 & 0 & 0 & 0 & 1 \\ 1 & 0 & 1 & 0 & 1 & 0 & 0 \\ 1 & 0 & 1 & 0 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 & 0 & 0 & 1 \\ 1 & 0 & 0 & 1 & 1 & 0 & 0 \\ 1 & 0 & 0 & 1 & 0 & 1 & 0 \\ 1 & 0 & 0 & 1 & 0 & 0 & 1 \end{pmatrix}, \text{ and } \beta = (\mu, \alpha_1, \alpha_2, \alpha_3, \beta_1, \beta_2, \beta_3)^\top.$$

(a) $\lambda_0\mu + \lambda_1\alpha_1 + \lambda_2\alpha_2 + \lambda_3\alpha_3 + \lambda_4\beta_1 + \lambda_5\beta_2 + \lambda_6\beta_3$ is estimable only when $\lambda^\top H = \lambda^\top$, where $H = (X^\top X)^{-1} X^\top X$.

(b) Not estimable.

(c) Not estimable.

(d) $\mu + \alpha_2$ is not estimable.

(e) $6\mu + 2\alpha_1 + 2\alpha_2 + 2\alpha_3 + 3\beta_1 + 3\beta_3$ is estimable.

(f) $\alpha - 2\alpha_2 + \alpha_3$ is estimable.

5. $Y = X\beta + \varepsilon$, where $X = \begin{pmatrix} 1 & 1 & 0 \\ 1 & 0 & 1 \\ 1 & 1 & 0 \end{pmatrix}$. $q^\top \beta$ is estimable if and only if $q^\top H = q^\top$, where $H = GX^\top X$.

$\lambda_1\beta_1 + \lambda_2\beta_2 + \lambda_3\beta_3 = q^\top \beta$ is estimable if and only if $q^\top H = q^\top$. Therefore $\lambda_1 = \lambda_2 + \lambda_3$.

6. $\beta = 2\pi\sqrt{1/g}$. $\hat{\beta} = \frac{\sum_{i=1}^k \sum_{j=1}^{n_i} \sqrt{l_i} t_{ij}}{\sum_{i=1}^k n_i l_i}$, and $Var(\hat{\beta}) = \frac{\sigma^2}{\sum_{i=1}^k n_i l_i}$.

7. (a) The null hypothesis

$$H_0 : \frac{\mu + \alpha_1}{a_1} = \frac{\mu + \alpha_2}{a_2} = \dots = \frac{\mu + \alpha_k}{a_k}$$

can also be presented as $K^\top \beta = m$, where $\beta = 0$ and $K = \begin{pmatrix} a_2 - a_1 & a_2 - a_1 & 0 & \dots & 0 \\ a_3 - a_2 & 0 & a_3 - a_2 & \dots & 0 \\ \vdots & & & & \\ a_k - a_{k-1} & 0 & 0 & \dots & a_k - a_{k-1} \end{pmatrix}$.

Since $H = GX^\top X = \begin{pmatrix} 0 & 0 \\ I_{1 \times k} & I_{k \times k} \end{pmatrix}$, then $K^\top H = K^\top$. Therefore, H_0 is testable.

(b) $F(H) \sim F_{k-1, n-k}$.

8. (a) $\beta = (\mu, \alpha_1, \alpha_2)^\top$ is not identifiable.

(b) β is identifiable.

(c) β is identifiable.

9. $A = E(\dot{g}(x)) = E(XX^\top \text{Dirac}(\varepsilon_\tau))$. $B = E\{XX^\top(\tau^2 1(\varepsilon_\tau > 0) + (\tau - 1)^2 1(\varepsilon_\tau \leq 0))\}$.