

1. With $E(Y_f) = H\beta \in R^q$, $LUP(Y_f) = LUE(H\beta) = [H(U^{1/2}X)^+U^{1/2} + \mathcal{N}(I_q, X)]Y$. Here $\mathcal{A} = H(U^{1/2}X)^+U^{1/2} + \mathcal{N}(I_q, X)$ is an affine set in $R^{q \times n}$.

- (1) Show that $H(U^{1/2}X)^+U^{1/2}XX^+ \in \mathcal{A}$.

Comment: Consequently $\mathcal{A} = H(U^{1/2}X)^+U^{1/2}XX^+ + \mathcal{N}(I_q, X)$.

$$\begin{aligned} \text{Write} \quad & H(U^{1/2}X)^+U^{1/2}XX^+ = H(U^{1/2}X)^+U^{1/2} + Z \\ \text{where} \quad & Z = H(U^{1/2}X)^+U^{1/2}(XX^+ - I). \\ \text{From} \quad & I_q ZX = I_q H(U^{1/2}X)^+U^{1/2}(X - X) = 0, \quad Z \in \mathcal{N}(I_q, X). \\ \text{Hence} \quad & H(U^{1/2}X)^+U^{1/2}XX^+ \in H(U^{1/2}X)^+U^{1/2} + \mathcal{N}(I_q, X) = \mathcal{A} \end{aligned}$$

- (2) Show that $H(U^{1/2}X)^+U^{1/2}XX^+ \perp \mathcal{N}(I_q, X)$.

Comment: Consequently, in \mathcal{A} , $H(U^{1/2}X)^+U^{1/2}XX^+$ has minimum norm.

For $Z \in \mathcal{N}(I_q, X)$,

$$\begin{aligned} \langle H(U^{1/2}X)^+U^{1/2}XX^+, Z \rangle &= \text{tr}(Z'H(U^{1/2}X)^+U^{1/2}XX^+) \\ &= \text{tr}(H(U^{1/2}X)^+U^{1/2}XX^+Z') \end{aligned}$$

$$\begin{aligned} \text{But} \quad & XX^+Z' = (X^+)'X'Z' = (X^+)'(I_q ZX)' = 0. \\ \text{So} \quad & \langle H(U^{1/2}X)^+U^{1/2}XX^+, Z \rangle = 0 \text{ for all } Z \in \mathcal{N}(I_q, X). \\ \text{Hence} \quad & H(U^{1/2}X)^+U^{1/2}XX^+ \perp \mathcal{N}(I_q, X). \end{aligned}$$

2. $\text{MSE}(\hat{u}, v) = E\|\hat{u} - v\|^2 = E[(\hat{u} - v)'(\hat{u} - v)]$ is a real-valued risk when v is predicted/estimated by $\hat{u} \in R^q$. The matrix-valued risk in the lecture is denoted as $\text{MSEM}(\hat{u}, v)$.

- (1) Suppose $q = 1$. Show that $\text{MSEM}(\hat{u}, v) = \text{MSE}(\hat{u}, v)$.

$$\begin{aligned} \text{When } q = 1, \text{MSEM}(\hat{u}, v) &= E[(\hat{u} - v)(\hat{u} - v)'] = E[(\hat{u} - v)(\hat{u} - v)] \\ &= E[(\hat{u} - v)'(\hat{u} - v)] = \text{MSE}(\hat{u}, v). \end{aligned}$$

- (2) Show that $\text{MSE}(\hat{u}, v) = \text{tr}[\text{MSEM}(\hat{u}, v)]$.

Hint: $E[\text{tr}(X)] = \text{tr}[E(X)]$.

$$\begin{aligned} \text{MSE}(\hat{u}, v) &= E[(\hat{u} - v)'(\hat{u} - v)] = E\{\text{tr}[(\hat{u} - v)'(\hat{u} - v)]\} = E\{\text{tr}[(\hat{u} - v)(\hat{u} - v)']\} \\ &= \text{tr}\{E[(\hat{u} - v)(\hat{u} - v)']\} = \text{tr}[\text{MSEM}(\hat{u}, v)]. \end{aligned}$$

- (3) Show that if \hat{u} dominates \tilde{u} by $\text{MSEM}(\cdot, \cdot)$, then \hat{u} dominates \tilde{u} by $\text{MSE}(\cdot, \cdot)$.

Hint: $A \leq 0 \implies \text{tr}(A) \leq 0$, and $A \geq 0 \implies \text{tr}(A) \geq 0$.

$$\begin{aligned} \hat{u} \text{ dominates } \tilde{u} \text{ by MSEM} &\implies \text{MSEM}(\tilde{u}, v) - \text{MSEM}(\hat{u}, v) \geq 0 \\ &\implies \text{tr}[\text{MSEM}(\tilde{u}, v) - \text{MSEM}(\hat{u}, v)] \geq 0 \\ &\implies \text{MSE}(\tilde{u}, v) - \text{MSE}(\hat{u}, v) \geq 0. \\ &\implies \hat{u} \text{ dominates } \tilde{u} \text{ by MSE.} \end{aligned}$$