

STAT-S632

Assignment 0

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Problem 1

Bias is defined as:

$$bias(\hat{y}|x) = (x_{1i}^T (X_1^T X_1)^{-1} X_1^T X_2^T - x_{2i}^T) \beta_2$$

Then:

$$\begin{aligned} \sum_{i=1}^n (bias(\hat{y}|x))^2 &= \sum \beta_2^T (x_{1i}^T (X_1^T X_1)^{-1} X_1^T X_2^T - x_{2i}^T)^T (x_{1i}^T (X_1^T X_1)^{-1} X_1^T X_2^T - x_{2i}^T) \beta_2 \\ &= \sum \beta_2^T (X_2^T X_1 (X_1^T X_1)^{-1} x_{1i} - x_{2i}) (x_{1i}^T (X_1^T X_1)^{-1} X_1^T X_2^T - x_{2i}^T) \beta_2 \end{aligned}$$

If we FOIL expand the expression:

$$= \sum \beta_2^T X_2^T (X_1 (X_1^T X_1)^{-1} x_{1i} x_{1i}^T (X_1^T X_1)^{-1} X_1^T) X_2 \beta_2 - 2 \sum \beta_2^T X_2^T X_1 (X_1^T X_1)^{-1} x_{1i} x_{2i}^T \beta_2 + \sum \beta_2^T x_{2i} x_{2i}^T \beta_2$$

Then using the definition of matrix multiplication:

$$= \beta_2^T X_2^T X_1 (X_1^T X_1)^{-1} X_1^T X_1 (X_1^T X_1)^{-1} X_1^T X_2 \beta_2 - 2 \beta_2^T X_2^T X_1 (X_1^T X_1)^{-1} X_1^T X_2 \beta_2 + \beta_2^T X_2^T X_2 \beta_2$$

The first term contains a $(X_1^T X_1)^{-1} (X_1^T X_1)$ which reduces to I :

$$= \beta_2^T X_2^T (X_1 I (X_1^T X_1)^{-1} X_1^T) X_2 \beta_2 - 2 \beta_2^T X_2^T (X_1 (X_1^T X_1)^{-1} X_1^T) X_2 \beta_2 + \beta_2^T X_2^T X_2 \beta_2$$

And we can recognize that $H_1 = X_1 (X_1^T X_1)^{-1} X_1^T$:

$$= (X_2 \beta_2)^T H_1 (X_2 \beta_2) - 2 (X_2 \beta_2)^T H_1 (X_2 \beta_2) + (X_2 \beta_2)^T I (X_2 \beta_2)$$

$$= (X_2 \beta_2)^T I (X_2 \beta_2) - (X_2 \beta_2)^T H_1 (X_2 \beta_2)$$

$$= (X_2 \beta_2)^T (I - H_1) (X_2 \beta_2)$$

Problem 2

Part a

Recall from S631 that $(H - H_1)$ is symmetric and idempotent.

Lemma: If matrix A is symmetric and idempotent, then it is positive semidefinite

Proof: Let z be a nonzero vector. Then since A is symmetric and idempotent,
 $z^T A z = z^T A A z = z^T A^T A z = (A z)^T (A z)$

Define $y = A z$. Then the above is $y^T y \geq 0$. Therefore, A is positive semidefinite.

Since $(H - H_1)$ is symmetric and idempotent, it is positive semidefinite.

Part b