

# 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

If  $z = e_i$  where  $e_i$  is a column vector such that the  $i^{\text{th}}$  element is 1 and the rest are 0, then for any matrix  $A$ ,  $z^T A z = [A]_{ii}$ . Then  $z^T (H - H_1) z = [H - H_1]_{ii} = h_{ii}^F - h_{ii}^R$ . Since  $H - H_1$  is positive semidefinite,  $z^T (H - H_1) z \geq 0$ . Therefore,  $h_{ii}^F - h_{ii}^R \geq 0 \implies h_{ii}^F \geq h_{ii}^R$ .