

# S631 HW5

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## 1. Show the following equalities:

a)  $\sum_{i=1}^n (y_i - x_i^T \hat{\beta})^2 = (\mathbf{Y} - \mathbf{X}\hat{\beta})^T (\mathbf{Y} - \mathbf{X}\hat{\beta})$

From lecture, we know that  $y_i$  represents a single observed value of a random variable, and there are 1:n of these observed values. Furthermore, we know that  $x_i^T$  is the transpose of all regressors used to construct a linear model in relation to our observed  $y_i$ , with 1:p different regressors for the 1:n observations. Finally,  $\hat{\beta}$  is a vector of unknown regression coefficients ( $\beta = \beta_0, \beta_1, \dots, \beta_p$ ) related to the regression of our  $x_i$ 's on  $y_i$ . With the summation of these terms, we are able to combine all of the individual values into matrices of dimension  $n \times 1$  (for  $\mathbf{Y}$ ),  $n \times (p+1)$  (for  $\mathbf{X}$ ), and  $(p+1) \times 1$  (for  $\beta$ ) as we are now dealing not with the individual values but their summations.

With this, we can then see that the RHS of this equality, when expressed with the dimensions of the matrices, becomes

$$(\mathbf{Y}_{n \times 1} - \mathbf{X}_{n \times (p+1)} \hat{\beta}_{(p+1) \times 1})^T (\mathbf{Y}_{n \times 1} - \mathbf{X}_{n \times (p+1)} \hat{\beta}_{(p+1) \times 1})$$

which when simplified becomes

$$(\mathbf{Y}_{1 \times n}^T - \hat{\beta}^T \mathbf{X}_{1 \times n}) (\mathbf{Y}_{n \times 1} - \mathbf{X}_{n \times 1} \hat{\beta})$$

and then

$$(\mathbf{Y}^T - \hat{\beta}^T \mathbf{X})_{1 \times n} (\mathbf{Y} - \mathbf{X} \hat{\beta})_{n \times 1}$$

From this we can see that the final product of the RHS is a  $1 \times 1$  scalar. This means that the final product of the LHS also must be a  $1 \times 1$  scalar.

Furthermore, from chapter 3.4.2 of ALR and our class notes, we know that  $y_i = x_i^T \beta + e_i$  and so,  $y_i - x_i^T \hat{\beta}$  is equal to the random vector of errors  $e_i$  and that the summation of these errors is a  $n \times 1$  matrix where  $\mathbf{e} = (e_1, \dots, e_n)^T$ .

This means

$$\sum_{i=1}^n (y_i - x_i^T \hat{\beta})^2$$

can be written as

$$\sum_{i=1}^n e_i^2 = \mathbf{e}^T \mathbf{e}$$

and in order to conclude with a  $1 \times 1$  scalar as the final product, we can then express this as

$$\mathbf{e}^2 = \mathbf{e}_{1 \times n}^T \mathbf{e}_{n \times 1} \quad \text{and so} \quad \mathbf{e}^T \mathbf{e} = (\mathbf{Y} - \mathbf{X} \hat{\beta})^T (\mathbf{Y} - \mathbf{X} \hat{\beta})$$

therefore

$$\sum_{i=1}^n (y_i - x_i^T \hat{\beta})^2 = (\mathbf{Y} - \mathbf{X} \hat{\beta})^T (\mathbf{Y} - \mathbf{X} \hat{\beta})$$

**b.  $H$  is symmetric and idempotent**

First, recall that  $H = X(X^T X)^{-1} X^T$ . Also, for  $H$  to be symmetric means  $H = H^T$ , and to be idempotent means  $HH = H$ .

Now, to show  $H$  is symmetric:

$$\begin{aligned}
 H^T &= H \\
 (X(X^T X)^{-1} X^T)^T &= X(X^T X)^{-1} X^T \\
 \text{proving the left side} \\
 X((X^T X)^{-1})^T X^T & \\
 X(X^{-1}(X^{-1})^T)^T X^T & \\
 X(X^{-1}(X^T)^{-1}) X^T & \\
 X(X^T X)^{-1} X^T &= H \\
 \blacksquare
 \end{aligned}$$

Now, to show  $H$  is idempotent:

$$\begin{aligned}
 H^2 &= H \\
 \text{proving the left side} \\
 (X(X^T X)^{-1} X_{p+1}^T)(X_{n \times p+1} (X^T X)^{-1} X^T) & \\
 X(X^T X)^{-1}_{p+1 \times p+1} (X^T X)_{p+1 \times p+1} (X^T X)^{-1} X^T & \\
 \text{using } AA^{-1} = I \text{ from matrix handout} & \\
 X_{n \times p+1} I_{p+1} (X^T X)^{-1} X^T & \\
 \text{using } A_{n \times m} I_m = A_{n \times m} & \\
 X(X^T X)^{-1} X^T &= H \\
 \blacksquare
 \end{aligned}$$

**c.  $I - H$  is symmetric and idempotent**

Now, to show  $I - H$  is symmetric:

$$\begin{aligned}
 (I - H)^T &= I - H \\
 I^T - H^T &= I - H \\
 \text{from part b we know } H^T &= H \\
 \text{we also know } I^T &= I \\
 \text{so } I - H &= I - H \\
 \blacksquare
 \end{aligned}$$

Now, to show  $I - H$  is idempotent:

$$(\mathbf{I} - \mathbf{H})^2 = \mathbf{I} - \mathbf{H}$$

proving the left side

$$(\mathbf{I} - \mathbf{H})(\mathbf{I} - \mathbf{H})$$

$$\mathbf{I}^2 - \mathbf{IH} - \mathbf{HI} + \mathbf{H}^2$$

we know  $\mathbf{I}^2 = \mathbf{I}$ ,  $\mathbf{IH}$  and  $\mathbf{HI} = \mathbf{H}$  as  $\mathbf{H}_n$  and  $\mathbf{I}_n$ , and from b  $\mathbf{H}^2 = \mathbf{H}$

so

$$\mathbf{I} - 2\mathbf{H} + \mathbf{H}$$

$$\mathbf{I} - \mathbf{H} = \mathbf{I} - \mathbf{H}$$

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d.  $\mathbf{HX} = \mathbf{X}$ .

First, recall from part b that  $\mathbf{AA}^{-1} = \mathbf{I}$  and  $\mathbf{A}_{n \times m} \mathbf{I}_m = \mathbf{A}_{n \times m}$ .

$$\mathbf{HX} = \mathbf{X}$$

proving left side

$$(\mathbf{X}(\mathbf{X}^T \mathbf{X})^{-1} \mathbf{X}^T) \mathbf{X}$$

$$\mathbf{X}(\mathbf{X}^T \mathbf{X})^{-1} (\mathbf{X}^T \mathbf{X})$$

$$\mathbf{XI} = \mathbf{X}$$

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e.  $(\mathbf{I} - \mathbf{H})(\mathbf{Y} - \mathbf{X}\hat{\boldsymbol{\beta}}) = (\mathbf{I} - \mathbf{H})\mathbf{Y}$

For this proof, first recall from class notes, and ALR chapter 9.1 that  $\mathbf{X}\hat{\boldsymbol{\beta}} = \hat{\mathbf{Y}}$ . Also recall from problem d that  $\mathbf{HX} = \mathbf{X}$ . Finally, recall again that  $\mathbf{A}_{n \times m} \mathbf{I}_m = \mathbf{A}_{n \times m}$ .

$$(\mathbf{I} - \mathbf{H})(\mathbf{Y} - \mathbf{X}\hat{\boldsymbol{\beta}}) = (\mathbf{I} - \mathbf{H})\mathbf{Y}$$

proving the left side

$$\mathbf{IY} - \mathbf{IX}\hat{\boldsymbol{\beta}} - \mathbf{HY} + \mathbf{HX}\hat{\boldsymbol{\beta}}$$

$$\mathbf{IY} - \mathbf{I}_n \hat{\mathbf{Y}}_{n \times 1} - \mathbf{HY} + \mathbf{X}\hat{\boldsymbol{\beta}}$$

$$\mathbf{IY} - \hat{\mathbf{Y}} - \mathbf{HY} + \hat{\mathbf{Y}}$$

$$\mathbf{IY} - \mathbf{HY}$$

$$(\mathbf{I} - \mathbf{H})\mathbf{Y}$$

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f.  $(\mathbf{Y} - \mathbf{X}\hat{\boldsymbol{\beta}})^T (\mathbf{I} - \mathbf{H})(\mathbf{Y} - \mathbf{X}\hat{\boldsymbol{\beta}}) = \mathbf{Y}^T (\mathbf{I} - \mathbf{H})\mathbf{Y}$

First, recall from problem d that  $\mathbf{HX} = \mathbf{X}$ , also from problem e that  $(\mathbf{I} - \mathbf{H})(\mathbf{Y} - \mathbf{X}\hat{\boldsymbol{\beta}}) = (\mathbf{I} - \mathbf{H})\mathbf{Y}$ , also from part b  $\mathbf{A}_{n \times m} \mathbf{I}_m = \mathbf{A}_{n \times m}$ , finally from ALR chapter 9.1, if  $\mathbf{X}\hat{\boldsymbol{\beta}} = \hat{\mathbf{Y}}$  then  $(\mathbf{X}\hat{\boldsymbol{\beta}})^T = \hat{\mathbf{Y}}^T$ .

$$\begin{aligned}
& (\mathbf{Y} - \mathbf{X}\hat{\boldsymbol{\beta}})^T(\mathbf{I} - \mathbf{H})(\mathbf{Y} - \mathbf{X}\hat{\boldsymbol{\beta}}) = \mathbf{Y}^T(\mathbf{I} - \mathbf{H})\mathbf{Y} \\
& \quad \text{proving LHS} \\
& (\mathbf{Y} - \mathbf{X}\hat{\boldsymbol{\beta}})^T(\mathbf{I} - \mathbf{H})\mathbf{Y} \\
& (\mathbf{Y}^T\mathbf{I} - \mathbf{Y}^T\mathbf{H} - \hat{\boldsymbol{\beta}}^T\mathbf{X}^T\mathbf{I} + \hat{\boldsymbol{\beta}}^T\mathbf{X}^T\mathbf{H})\mathbf{Y} \\
& (\mathbf{Y}^T\mathbf{I} - \mathbf{Y}^T\mathbf{H} - \hat{\mathbf{Y}}_{1 \times n}^T\mathbf{I}_n + \hat{\boldsymbol{\beta}}^T\mathbf{X}^T)\mathbf{Y} \\
& (\mathbf{Y}^T\mathbf{I} - \mathbf{Y}^T\mathbf{H} - \hat{\mathbf{Y}}^T + \hat{\mathbf{Y}}^T)\mathbf{Y} \\
& \quad \mathbf{Y}^T(\mathbf{I} - \mathbf{H})\mathbf{Y} \\
& \quad \blacksquare
\end{aligned}$$

g.  $RSS(\hat{\boldsymbol{\beta}}) = \mathbf{Y}^T(\mathbf{I} - \mathbf{H})\mathbf{Y}$

For this proof, recall that  $RSS(\hat{\boldsymbol{\beta}}) = (\mathbf{Y} - \mathbf{X}\hat{\boldsymbol{\beta}})^T(\mathbf{Y} - \mathbf{X}\hat{\boldsymbol{\beta}})$ , also from ALR 9.1 that  $\mathbf{Y} - \hat{\mathbf{Y}} = (\mathbf{I} - \mathbf{H})\mathbf{Y}$ , from problem c that  $(\mathbf{I} - \mathbf{H})$  is symmetric and idempotent, and finally from problem f that  $\mathbf{X}\hat{\boldsymbol{\beta}} = \hat{\mathbf{Y}}$ .

$$\begin{aligned}
& RSS(\hat{\boldsymbol{\beta}}) = \mathbf{Y}^T(\mathbf{I} - \mathbf{H})\mathbf{Y} \\
& \quad \text{proving LHS} \\
& (\mathbf{Y} - \mathbf{X}\hat{\boldsymbol{\beta}})^T(\mathbf{Y} - \mathbf{X}\hat{\boldsymbol{\beta}}) \\
& ((\mathbf{I} - \mathbf{H})\mathbf{Y})^T(\mathbf{I} - \mathbf{H})\mathbf{Y} \\
& \mathbf{Y}^T(\mathbf{I} - \mathbf{H})^T(\mathbf{I} - \mathbf{H})\mathbf{Y} \\
& \mathbf{Y}^T(\mathbf{I} - \mathbf{H})(\mathbf{I} - \mathbf{H})\mathbf{Y} \\
& \mathbf{Y}^T(\mathbf{I} - \mathbf{H})\mathbf{Y} \\
& \quad \blacksquare
\end{aligned}$$