

# The Orthogonal Projection Matrix onto $\mathcal{C}(\mathbf{X})$ and the Normal Equations

## Theorem 2.1:

$P_X$ , the orthogonal projection matrix onto  $\mathcal{C}(X)$ , is equal to  $X(X'X)^{-1}X'$ .

The matrix  $X(X'X)^{-1}X'$  satisfies

(a)  $[X(X'X)^{-1}X'] [X(X'X)^{-1}X'] = [X(X'X)^{-1}X']$

(b)  $X(X'X)^{-1}X'y \in \mathcal{C}(X) \forall y \in \mathbb{R}^n$

(c)  $X(X'X)^{-1}X'x = x \forall x \in \mathcal{C}(X)$

(d)  $X(X'X)^{-1}X' = [X(X'X)^{-1}X']'$

(e)  $X(X'X)^{-1}X'$  is the same  $\forall$  GI of  $X'X$ .

It is useful to establish a few results before proving Theorem 2.1.

Show that if  $\mathbf{A}$  is a symmetric matrix and  $\mathbf{G}$  is a GI of  $\mathbf{A}$ , then  $\mathbf{G}'$  is also GI of  $\mathbf{A}$ .

## Result 2.4:

$$\mathbf{X}'\mathbf{X}\mathbf{A} = \mathbf{X}'\mathbf{X}\mathbf{B} \iff \mathbf{X}\mathbf{A} = \mathbf{X}\mathbf{B}.$$

One direction is obvious ( $\Leftarrow$ ).

Can you prove the other ( $\Rightarrow$ )?

## Result 2.5:

Suppose  $(X'X)^-$  is any GI of  $X'X$ . Then  $(X'X)^-X'$  is a GI of  $X$ , i.e.,

$$X(X'X)^-X'X = X.$$

## Corollary to Result 2.5:

For  $(X'X)^-$  any GI of  $X'X$ ,

$$X'X(X'X)^-X' = X'.$$

## Now we can prove Theorem 2.1:

First show  $\mathbf{X}(\mathbf{X}'\mathbf{X})^{-1}\mathbf{X}'$  is idempotent.

Now show that

$$\mathbf{X}(\mathbf{X}'\mathbf{X})^{-1}\mathbf{X}'\mathbf{y} \in \mathcal{C}(\mathbf{X}) \quad \forall \mathbf{y} \in \mathbb{R}^n.$$



Now show that

$$\mathbf{X}(\mathbf{X}'\mathbf{X})^{-1}\mathbf{X}'\mathbf{z} = \mathbf{z} \quad \forall \mathbf{z} \in \mathcal{C}(\mathbf{X}).$$

Now we know  $X(X'X)^{-1}X'$  is a projection matrix onto  $\mathcal{C}(X)$ .

To be the unique projection matrix onto  $\mathcal{C}(X)$ ,  $X(X'X)^{-1}X'$  must be symmetric.

To show symmetry of  $X(X'X)^-X'$ , it will help to first show that

$$X(X'X)_1^-X' = X(X'X)_2^-X'$$

for any two GIs of  $X'X$  denoted  $(X'X)_1^-$  and  $(X'X)_2^-$ .

Now show that  $\mathbf{X}(\mathbf{X}'\mathbf{X})^{-1}\mathbf{X}'$  is symmetric.

We have shown that  $\mathbf{X}(\mathbf{X}'\mathbf{X})^{-1}\mathbf{X}'$  is the orthogonal projection matrix onto  $\mathcal{C}(\mathbf{X})$ .

We know that it is the only symmetric projection matrix onto  $\mathcal{C}(\mathbf{X})$  by Result A.16.

## Example:

Find the orthogonal projection matrix onto  $\mathcal{C}(\mathbf{1}_{n \times 1})$ .

Also, simplify  $P_X y$  in this case.

By Corollary A.4 to Result A.16, we know that

$$\mathbf{I} - \mathbf{P}_X = \mathbf{I} - \mathbf{X}(\mathbf{X}'\mathbf{X})^{-1}\mathbf{X}'$$

is the orthogonal projection matrix onto  $\mathcal{C}(\mathbf{X})^\perp = \mathcal{N}(\mathbf{X}')$ .

This is Result 2.6.

By Result A.4, any  $\mathbf{y} \in \mathbb{R}^n$  may be written as

$$\mathbf{y} = \mathbf{s} + \mathbf{t},$$

where  $\mathbf{s} \in \mathcal{C}(\mathbf{X})$  and  $\mathbf{t} \in \mathcal{C}(\mathbf{X})^\perp = \mathcal{N}(\mathbf{X}')$ .

Furthermore, the vectors  $\mathbf{s} \in \mathcal{C}(\mathbf{X})$  and  $\mathbf{t} \in \mathcal{C}(\mathbf{X})^\perp$  are unique.



Because

$$\mathbf{y} = \mathbf{P}_X \mathbf{y} + (\mathbf{I} - \mathbf{P}_X) \mathbf{y}$$

$$\mathbf{P}_X \mathbf{y} \in \mathcal{C}(X) \text{ and } (\mathbf{I} - \mathbf{P}_X) \mathbf{y} \in \mathcal{C}(X)^\perp,$$

we get  $s = \mathbf{P}_X \mathbf{y}$  and  $t = (\mathbf{I} - \mathbf{P}_X) \mathbf{y}$ .

Moreover,

$$\hat{\mathbf{y}} \equiv \mathbf{P}_X \mathbf{y} = \text{the vector of fitted values}$$

$$\hat{\boldsymbol{\varepsilon}} \equiv (\mathbf{I} - \mathbf{P}_X) \mathbf{y} = \mathbf{y} - \hat{\mathbf{y}} = \text{the vector of residuals.}$$

Let  $P_W$  and  $P_X$  denote the orthogonal projection matrices onto  $\mathcal{C}(W)$  and  $\mathcal{C}(X)$ , respectively.

Suppose  $\mathcal{C}(W) \subseteq \mathcal{C}(X)$ . Show that

$$P_W P_X = P_X P_W = P_W.$$

## Theorem 2.2:

If  $\mathcal{C}(\mathbf{W}) \subseteq \mathcal{C}(\mathbf{X})$ , then  $\mathbf{P}_X - \mathbf{P}_W$  is the orthogonal projection matrix onto  $\mathcal{C}((\mathbf{I} - \mathbf{P}_W)\mathbf{X})$ .

We have previously seen that

$$\begin{aligned} Q(\mathbf{b}) &= (\mathbf{y} - \mathbf{X}\mathbf{b})'(\mathbf{y} - \mathbf{X}\mathbf{b}) \\ &= \|\mathbf{y} - \mathbf{X}\mathbf{b}\|^2 \geq \|\mathbf{y} - \mathbf{P}_X\mathbf{y}\|^2 \end{aligned}$$

$\forall \mathbf{b} \in \mathbb{R}^p$  with equality iff  $\mathbf{X}\mathbf{b} = \mathbf{P}_X\mathbf{y}$ .

Now we know that  $P_X = X(X'X)^{-1}X'$ .

Thus,  $\hat{\beta}$  minimizes  $Q(\mathbf{b})$  iff  $X\hat{\beta} = X(X'X)^{-1}X'y$ .

By Result 2.4, this equation is equivalent to

$$X'X\hat{\beta} = X'X(X'X)^{-1}X'y.$$

Because  $X'X(X'X)^{-1}X' = X'$ ,  $X'X\hat{\beta} = X'X(X'X)^{-1}X'y$  is equivalent to

$$X'X\hat{\beta} = X'y.$$

This system of linear equations is known as the Normal Equations (NE).

We have established Result 2.3:

$\hat{\beta}$  is a solution to the NE ( $X'X\mathbf{b} = X'\mathbf{y}$ ) iff  $\hat{\beta}$  minimizes  $Q(\mathbf{b})$ .

## Corollary 2.1:

The NE are consistent.



By Result A.13,  $\hat{\beta}$  is a solution to  $X'Xb = X'y$  iff

$$\hat{\beta} = (X'X)^{-1}X'y + [I - (X'X)^{-1}X'X]z$$

for some  $z \in \mathbb{R}^p$ .

## Corollary 2.3:

$X\hat{\beta}$  is invariant to the choice of a solution  $\hat{\beta}$  to the NE, i.e., if  $\hat{\beta}_1$  and  $\hat{\beta}_2$  are any two solutions to the NE, then  $X\hat{\beta}_1 = X\hat{\beta}_2$ .

We will finish this set of notes with some other results from Section 2.2 of the text.

Lemma 2.1:  $\mathcal{N}(\mathbf{X}'\mathbf{X}) = \mathcal{N}(\mathbf{X})$ .

Result 2.2:  $\mathcal{C}(\mathbf{X}'\mathbf{X}) = \mathcal{C}(\mathbf{X}')$ .

Corollary 2.2:  $\text{rank}(\mathbf{X}'\mathbf{X}) = \text{rank}(\mathbf{X})$ .