

Equivalence of the Full Model/Reduced Model and $H_0: \mathbf{C}\boldsymbol{\beta} = \mathbf{0}$ Sums of Squares

In a regression context, consider a (full rank model matrix)

$$\mathbf{X} = (\mathbf{1} | \mathbf{x}_1 | \mathbf{x}_2 | \dots | \mathbf{x}_r)$$

$n \times (r+1)$

For $p < r$, let

$$\mathbf{X}_p = (\mathbf{1} | \mathbf{x}_1 | \mathbf{x}_2 | \dots | \mathbf{x}_p)$$

and

$$\mathbf{U}_p = (\mathbf{x}_{p+1} | \mathbf{x}_{p+2} | \dots | \mathbf{x}_r)$$

For

$$\mathbf{C} = \left(\begin{array}{c|c} \mathbf{0} & \mathbf{I} \\ \hline (r-p) \times (p+1) & (r-p) \times (r-p) \end{array} \right)$$

consider

$$SS_{H_0} = (\mathbf{C}\mathbf{b}_{OLS})' (\mathbf{C}(\mathbf{X}'\mathbf{X})^{-1}\mathbf{C}') (\mathbf{C}\mathbf{b}_{OLS})$$

which we invented for testing $H_0: \mathbf{C}\boldsymbol{\beta} = \mathbf{0}$. The object here is to show that

$$\begin{aligned} SS_{H_0} &= SSE_{\text{Reduced}} - SSE_{\text{Full}} \\ &= \mathbf{Y}'(\mathbf{I} - \mathbf{P}_{\mathbf{X}_p})\mathbf{Y} - \mathbf{Y}'(\mathbf{I} - \mathbf{P}_{\mathbf{X}})\mathbf{Y} \\ &= \mathbf{Y}'(\mathbf{P}_{\mathbf{X}} - \mathbf{P}_{\mathbf{X}_p})\mathbf{Y} \end{aligned}$$

To begin, note that with

$$\mathbf{A} = (\mathbf{U}_p'(\mathbf{I} - \mathbf{P}_{\mathbf{X}_p})\mathbf{U}_p)^{-1}\mathbf{U}_p'(\mathbf{I} - \mathbf{P}_{\mathbf{X}_p})$$

we have

$$\mathbf{C} = \mathbf{A}\mathbf{X}$$

Then,

$$\begin{aligned} SS_{H_0} &= (\mathbf{A}\hat{\mathbf{Y}})'(\mathbf{A}\mathbf{P}_{\mathbf{X}}\mathbf{A}')^{-1}(\mathbf{A}\hat{\mathbf{Y}}) \\ &= \mathbf{Y}'\mathbf{P}_{\mathbf{X}}\mathbf{A}'(\mathbf{A}\mathbf{P}_{\mathbf{X}}\mathbf{P}_{\mathbf{X}}\mathbf{A}')^{-1}\mathbf{A}\mathbf{P}_{\mathbf{X}}\mathbf{Y} \end{aligned}$$

Then write

$$\begin{aligned} \mathbf{P}_{\mathbf{X}}\mathbf{A}' &= \mathbf{P}_{\mathbf{X}}(\mathbf{I} - \mathbf{P}_{\mathbf{X}_p})\mathbf{U}_p(\mathbf{U}_p'(\mathbf{I} - \mathbf{P}_{\mathbf{X}_p})\mathbf{U}_p)^{-1} \\ &= (\mathbf{U}_p - \mathbf{P}_{\mathbf{X}_p}\mathbf{U}_p)(\mathbf{U}_p'(\mathbf{I} - \mathbf{P}_{\mathbf{X}_p})(\mathbf{I} - \mathbf{P}_{\mathbf{X}_p})\mathbf{U}_p)^{-1} \\ &= (\mathbf{U}_p - \mathbf{P}_{\mathbf{X}_p}\mathbf{U}_p)\left((\mathbf{U}_p - \mathbf{P}_{\mathbf{X}_p}\mathbf{U}_p)'(\mathbf{U}_p - \mathbf{P}_{\mathbf{X}_p}\mathbf{U}_p)\right)^{-1} \end{aligned}$$

Use the abbreviation

$$\mathbf{U}_p^* = \mathbf{U}_p - \mathbf{P}_{\mathbf{X}_p}\mathbf{U}_p$$

and note that

$$\begin{aligned} SS_{H_0} &= \mathbf{Y}'\mathbf{U}_p^* (\mathbf{U}_p^{*\prime}\mathbf{U}_p^*)^{-1} \left((\mathbf{U}_p^{*\prime}\mathbf{U}_p^*)^{-1} \mathbf{U}_p^{*\prime}\mathbf{U}_p^* (\mathbf{U}_p^{*\prime}\mathbf{U}_p^*)^{-1} \right)^{-1} (\mathbf{U}_p^{*\prime}\mathbf{U}_p^*)^{-1} \mathbf{U}_p^{*\prime}\mathbf{Y} \\ &= \mathbf{Y}'\mathbf{P}_{\mathbf{U}_p^*}\mathbf{Y} \end{aligned}$$

It then suffices to show that $\mathbf{P}_{\mathbf{U}_p^*} = \mathbf{P}_{\mathbf{X}} - \mathbf{P}_{\mathbf{X}_p}$.

Christensen's Theorem B.47 shows that $\mathbf{P}_{\mathbf{X}} - \mathbf{P}_{\mathbf{X}_p}$ is a perpendicular projection matrix. His Theorem B.48 says that $C(\mathbf{P}_{\mathbf{X}} - \mathbf{P}_{\mathbf{X}_p})$ is the "orthogonal complement of $C(\mathbf{X}_p)$ with respect to $C(\mathbf{X})$ " defined on his page 395. That is,

$$C(\mathbf{P}_{\mathbf{X}} - \mathbf{P}_{\mathbf{X}_p}) = C(\mathbf{X}) \cap C(\mathbf{X}_p)^\perp$$

so it suffices to show that $C(\mathbf{U}_p^*) = C(\mathbf{X}) \cap C(\mathbf{X}_p)^\perp$.

Each column of $\mathbf{U}_p - \mathbf{P}_{\mathbf{X}_p}\mathbf{U}_p$ is a difference in a column of \mathbf{U}_p and a linear combination of columns of $\mathbf{P}_{\mathbf{X}_p}$. Since $C(\mathbf{U}_p) \subset C(\mathbf{X})$ and $C(\mathbf{P}_{\mathbf{X}_p}) = C(\mathbf{X}_p) \subset C(\mathbf{X})$, each column of $\mathbf{U}_p - \mathbf{P}_{\mathbf{X}_p}\mathbf{U}_p$ is in $C(\mathbf{X})$ and $C(\mathbf{U}_p^*) \subset C(\mathbf{X})$.

So then note that if $\mathbf{v} \in C(\mathbf{U}_p^*) \exists \gamma$ such that

$$\mathbf{v} = \mathbf{U}_p^* \gamma$$

For such a \mathbf{v}

$$(\mathbf{I} - \mathbf{P}_{\mathbf{X}_p})\mathbf{v} = (\mathbf{I} - \mathbf{P}_{\mathbf{X}_p})\mathbf{U}_p^* \gamma = (\mathbf{I} - \mathbf{P}_{\mathbf{X}_p})(\mathbf{I} - \mathbf{P}_{\mathbf{X}_p})\mathbf{U}_p \gamma = (\mathbf{I} - \mathbf{P}_{\mathbf{X}_p})\mathbf{U}_p \gamma = \mathbf{U}_p^* \gamma = \mathbf{v}$$

so $\mathbf{v} \in C(\mathbf{X}_p)^\perp$. Thus $C(\mathbf{U}_p^*) \subset C(\mathbf{X}) \cap C(\mathbf{X}_p)^\perp$.

Finally, suppose that $\mathbf{v} \in C(\mathbf{X}) \cap C(\mathbf{X}_p)^\perp$ and for some

$$\gamma = \begin{pmatrix} \gamma_1 \\ (p+1) \times 1 \\ \gamma_2 \\ (r-p) \times 1 \end{pmatrix}$$

write

$$\mathbf{v} = \mathbf{X}\gamma = (\mathbf{X}_p | \mathbf{U}_p)\gamma = (\mathbf{X}_p | \mathbf{U}_p^* + \mathbf{P}_{\mathbf{X}_p}\mathbf{U}_p)\gamma = \mathbf{X}_p\gamma_1 + (\mathbf{U}_p^* + \mathbf{P}_{\mathbf{X}_p}\mathbf{U}_p)\gamma_2$$

Then continuing

$$\mathbf{v} = \mathbf{X}_p\gamma_1 + \mathbf{P}_{\mathbf{X}_p}\mathbf{U}_p + \mathbf{U}_p^*\gamma_2 = \mathbf{P}_{\mathbf{X}_p}\mathbf{X}_p\gamma_1 + \mathbf{P}_{\mathbf{X}_p}\mathbf{U}_p + \mathbf{U}_p^*\gamma_2$$

so

$$\mathbf{v} = \mathbf{P}_{\mathbf{X}_p}(\mathbf{X}_p | \mathbf{U}_p)\gamma + \mathbf{U}_p^*\gamma_2 = \mathbf{P}_{\mathbf{X}_p}\mathbf{v} + \mathbf{U}_p^*\gamma_2$$

Since $\mathbf{v} \perp C(\mathbf{X}_p)$ we then have $\mathbf{v} = \mathbf{U}_p^*\gamma_2$ and thus $\mathbf{v} \in C(\mathbf{U}_p^*)$. So $C(\mathbf{X}) \cap C(\mathbf{X}_p)^\perp \subset C(\mathbf{U}_p^*)$ and we're done.