

Econometrics . 2022-4-15

## Partitioned Regression (cont.)

$$y = X\beta + \varepsilon$$

$$\begin{aligned} X &= [X_1 \ X_2] \Rightarrow y = [X_1 \ X_2] \begin{bmatrix} \beta_1 \\ \beta_2 \end{bmatrix} + \varepsilon \\ \text{(n \times k)} \quad \text{n \times s} \quad \text{n \times (k-s)} & \end{aligned}$$

long regression

$$= X_1\beta_1 + X_2\beta_2 + \varepsilon$$

The normal equations :

$$X'X\beta = X'y \Rightarrow \begin{bmatrix} X_1'X_1 & X_1'X_2 \\ X_2'X_1 & X_2'X_2 \end{bmatrix} \begin{bmatrix} \beta_1 \\ \beta_2 \end{bmatrix} = \begin{bmatrix} X_1'y \\ X_2'y \end{bmatrix} \begin{matrix} \text{--- ①} \\ \text{--- ②} \end{matrix}$$

(i)  $X_1$  and  $X_2$  are orthogonal.

$$X_1'X_2 = 0, \quad X_2'X_1 = 0.$$

$$\Rightarrow \begin{bmatrix} X_1'X_1 & 0 \\ 0 & X_2'X_2 \end{bmatrix} \begin{bmatrix} b_1 \\ b_2 \end{bmatrix} = \begin{bmatrix} X_1'y \\ X_2'y \end{bmatrix}.$$

$$\Leftrightarrow \begin{cases} X_1'X_1 b_1 = X_1'y \\ X_2'X_2 b_2 = X_2'y \end{cases} \Rightarrow \begin{cases} b_1 = (X_1'X_1)^{-1} X_1'y \\ b_2 = (X_2'X_2)^{-1} X_2'y \end{cases}.$$

$$y = X_1 \beta_1^* + \varepsilon_1^*$$

$$y = X_2 \beta_2^* + \varepsilon_2^*$$

short regressing

$a'b = 0 \Leftrightarrow b'a = 0$   
 $a$  and  $b$  are orthogonal

Homework

$x_1 \neq c x_2$ .

$$\begin{bmatrix} 1 \\ 1 \end{bmatrix} \quad \begin{bmatrix} 1 \\ 2 \end{bmatrix}$$

(ii)  $X_1$  and  $X_2$  are not orthogonal.

$$\textcircled{1} \Rightarrow b_1 = (X_1'X_1)^{-1}X_1'y - (X_1'X_1)^{-1}X_1'X_2b_2$$

Substitute it into  $\textcircled{2}$ .

$$\boxed{X_2'X_1b_1} + X_2'X_2b_2 = X_2'y$$

$$X_2'X_1(X_1'X_1)^{-1}X_1'y - X_2'X_1(X_1'X_1)^{-1}X_1'X_2b_2 + X_2'X_2b_2 = X_2'y$$

$$\underbrace{X_2'X_2}_{\text{I}} - \underbrace{X_2'X_1(X_1'X_1)^{-1}X_1'X_2}_{\text{I}}]b_2 = [X_2' - X_2'X_1(X_1'X_1)^{-1}X_1']y$$

$$X_2' \underbrace{[I - X_1(X_1'X_1)^{-1}X_1']}_{\text{I}} X_2 b_2 = X_2' \underbrace{[I - X_1(X_1'X_1)^{-1}X_1']}_{\text{I}} y$$

Residual maker:  $M = I - X(X'X)^{-1}X'$ . ( $y = X\beta + \varepsilon$ )

From the short regression  $y = X_1\beta_1^* + \varepsilon_1^*$ .

$$M_1 = I - X_1(X_1'X_1)^{-1}X_1'$$

$$\Rightarrow \boxed{X_2'M_1X_2\beta_2 = X_2'M_1y}$$

$$\begin{aligned}\Rightarrow \beta_2 &= (X_2'M_1X_2)^{-1}(X_2'M_1y) \\ &= (X_2'M_1' M_1 X_2)^{-1}(X_2'M_1' M_1 y) \\ &= \left[ \underbrace{(M_1 X_2)'} \underbrace{M_1 X_2} \right]^{-1} \underbrace{(M_1 X_2)'} \underbrace{M_1 y}\end{aligned}$$

residual.

$M_1$  is symmetric  
and idempotent.

$$\begin{aligned}M_1 &= M_1 \cdot M_1 \\ &= M_1' \cdot M_1\end{aligned}$$

$$\begin{aligned}
 M_1 X_2 &= M_1 [x_{s+1} \ x_{s+2} \ \dots \ x_k] \\
 &= [M_1 x_{s+1} \ M_1 x_{s+2} \ \dots \ M_1 x_k]
 \end{aligned}$$

$$\left[ \begin{array}{l}
 x_{s+1} = X_1 \beta_{s+1}^* + \varepsilon_{s+1}^* \xrightarrow{\text{residual}} M_1 x_{s+1}. \\
 x_{s+2} = X_1 \beta_{s+2}^* + \varepsilon_{s+2}^* \xrightarrow{\hspace{1cm}} M_1 x_{s+2} \\
 \vdots \\
 x_k = X_1 \beta_k^* + \varepsilon_{s+2}^* \xrightarrow{\hspace{1cm}} M_1 x_k.
 \end{array} \right.$$

Let  $X_2^* = M_1 X_2$  ,  $y^* = M_1 y$  , we have

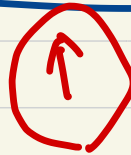
$$b_2 = (X_2^{*'} X_2^*)^{-1} X_2^{*'} y^* .$$

OLS solution  
 $b = (X'X)^{-1} X'y$

$b_2$  is the OLS solution of

$$y^* = X_2^* \beta_2^{**} + \varepsilon_2^{**}$$

↑  
coefficient



# [Frisch-Waugh-Lovell Theorem]

In the linear regression  $y = X_1\beta_1 + X_2\beta_2 + \varepsilon$ ,  
the least squares solution  $\beta_2$  is the set of coefficients  
obtained when the residual from a regression of  $y$  on  $X_1$   
alone are regressed on the set of residuals obtained " $y^*$ "  
when each column of  $X_2$  is regressed on  $X_1$  " $X_2^*$ "

The FWL Theorem says that the fitted values  $\hat{b}_2$  in the following two models are identical.

(S1)  $y = X_1\beta + X_2\beta_2 + \varepsilon$  (the long regression)

$\Rightarrow$  OLS coefficients  $\hat{b}_2 = (X_2' M_1 X_2)^{-1} (X_2' M_1 y)$ .

(S2) Step 1. Fit the models and calculate the residuals

$y = X_1\beta_1^* + \varepsilon_1^* \Rightarrow e_y = M_1 y$

columns  
in  $X_2$

$\left\{ \begin{array}{l} x_{s+1} = X_1\beta_{s+1}^* + \varepsilon_{s+1}^* \\ \vdots \\ x_k = X_1\beta_k^* + \varepsilon_k^* \end{array} \right.$

$e_{s+1} = M_1 x_{s+1}$

$\vdots$   
 $e_k = M_1 x_k$

$\hat{b}_2^* = (X_2^{*'} X_2^*)^{-1} X_2^{*'} e_y^*$

Step 2. Put  $y^* = e_y$ ,  $X_2^* = [e_{s+1} \dots e_k]$

$\Rightarrow$  The OLS coefficients of  $y^* = X_2^* \beta_2^{*'} + \varepsilon_1^*$   
 $\Rightarrow \hat{b}_2^* = \hat{b}_2$