

Enrollment Key 1341

EE 435

$$Y_i = \beta_1 + \beta_2 X_{2i} + \dots + \beta_k X_{ki} + u_i \quad i=1, 2, \dots, n$$

$$\begin{matrix} i=1 \\ i=2 \\ \vdots \\ i=n \end{matrix} \begin{bmatrix} Y_1 \\ Y_2 \\ \vdots \\ Y_n \end{bmatrix} = \begin{bmatrix} \beta_1 + \beta_2 X_{21} + \dots + \beta_k X_{k1} \\ \beta_1 + \beta_2 X_{22} + \dots + \beta_k X_{k2} \\ \vdots \\ \beta_1 + \beta_2 X_{2n} + \dots + \beta_k X_{kn} \end{bmatrix} + \begin{bmatrix} u_1 \\ u_2 \\ \vdots \\ u_n \end{bmatrix}$$

$n \times 1 \qquad \qquad \qquad n \times 1 \qquad \qquad \qquad n \times 1$

$$\begin{bmatrix} Y_1 \\ Y_2 \\ \vdots \\ Y_n \end{bmatrix}_{n \times 1} = \begin{bmatrix} 1 & X_{21} & \dots & X_{k1} \\ 1 & X_{22} & \dots & X_{k2} \\ \vdots & \vdots & \ddots & \vdots \\ 1 & X_{2n} & \dots & X_{kn} \end{bmatrix}_{n \times k} \cdot \begin{bmatrix} \beta_1 \\ \beta_2 \\ \vdots \\ \beta_k \end{bmatrix}_{k \times 1} + \begin{bmatrix} u_1 \\ u_2 \\ \vdots \\ u_n \end{bmatrix}_{n \times 1}$$

$$\boxed{Y = X\beta + u}$$

$n \times 1 \qquad \qquad \qquad n \times k \quad k \times 1 \quad n \times 1$

$$u = \begin{bmatrix} u_1 \\ u_2 \\ \vdots \\ u_n \end{bmatrix}$$

$$u' = [u_1 \ u_2 \ \dots \ u_n]$$

$$\hat{u}'\hat{u} = [\hat{u}_1 \ \hat{u}_2 \ \dots \ \hat{u}_n]_{1 \times n} \begin{bmatrix} \hat{u}_1 \\ \hat{u}_2 \\ \vdots \\ \hat{u}_n \end{bmatrix}_{n \times 1} = \hat{u}_1^2 + \hat{u}_2^2 + \dots + \hat{u}_n^2 = \sum_{i=1}^n \hat{u}_i^2$$

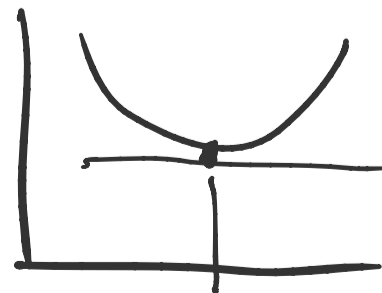
$$Y = X\hat{\beta} + \hat{u} \Rightarrow \hat{u} = (Y - X\hat{\beta})$$

$$\hat{u}'\hat{u} = (Y - X\hat{\beta})'(Y - X\hat{\beta})$$

$$\hat{u}'\hat{u} = Y'Y - Y'X\hat{\beta} - \hat{\beta}'X'Y + \hat{\beta}'X'X\hat{\beta}$$

$(1 \times n) \quad (n \times k)$ $(1 \times k) \quad (k \times n)$ $(1 \times k)$ $(k \times n)$ $(n \times 1)$

$$\frac{\partial \hat{u}'\hat{u}}{\partial \hat{\beta}} = \frac{\partial Y'Y}{\partial \hat{\beta}} - \frac{\partial Y'X\hat{\beta}}{\partial \hat{\beta}} - \frac{\partial \hat{\beta}'X'Y}{\partial \hat{\beta}} + \frac{\partial \hat{\beta}'X'X\hat{\beta}}{\partial \hat{\beta}} = 0$$



$0 \quad -2Y'X \quad + \quad 2X'X\hat{\beta} \quad - \quad 0$

$$\ddot{0} - 2X'Y + 2X'X\hat{\beta} = 0$$

$$2X'X\hat{\beta} = 2X'Y$$

$$(\cancel{X'X})^{-1}X'X\hat{\beta} = (X'X)^{-1}X'Y$$

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

$\begin{matrix} k \times 1 & k \times n & n \times k & k \times n & n \times 1 \end{matrix}$

$$E(u \cdot u') = \begin{bmatrix} E(u_1 u_1) & E(u_1 u_2) & \dots & E(u_1 u_n) \\ E(u_2 u_1) & E(u_2 u_2) & \dots & E(u_2 u_n) \\ \vdots & \vdots & \ddots & \vdots \\ E(u_n u_1) & E(u_n u_2) & \dots & E(u_n u_n) \end{bmatrix} = \begin{bmatrix} \sigma_1^2 & \sigma_{12} & \dots & \sigma_{1n} \\ \sigma_{21} & \sigma_2^2 & \dots & \sigma_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ \sigma_{n1} & \sigma_{n2} & \dots & \sigma_n^2 \end{bmatrix}$$

$$E[(x_i - E(x_i))^2] = \text{var}(x_i)$$

$$E[(u_i - E(u_i))^2] = E(u_i^2) = E(u_1 u_1) = \text{var}(u_1) = \sigma_1^2$$

$$E[(x_i - E(x_i))(y_i - E(y_i))] = \text{cov}(x_i, y_i)$$

$$E[(u_1 - E(u_1))(u_2 - E(u_2))] = E(u_1 u_2) = \text{cov}(u_1, u_2) = \sigma_{12}$$

$$E(u \cdot u') = \sum_{i=1}^n \begin{bmatrix} \sigma_1^2 & \sigma_{12} & \dots & \sigma_{1n} \\ \sigma_{21} & \sigma_2^2 & \dots & \sigma_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ \sigma_{n1} & \sigma_{n2} & \dots & \sigma_n^2 \end{bmatrix} = \begin{bmatrix} \sigma^2 & 0 & \dots & 0 \\ 0 & \sigma^2 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & \sigma^2 \end{bmatrix} = \sigma^2 I_{n \times n}$$

Assume Homo or Non auto
 Cross-section Time-Series

$$\sigma_1^2 = \sigma_2^2 = \dots = \sigma_n^2 = \sigma^2 \quad \text{cov}(u_i, u_j) = 0$$

constant var.

Generalized Least Squares (GLS) Relax - (~~Homo~~
~~Non auto~~)

OLS
 (Homo & Non auto)

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

$$\sum_{n \times n}^1 = \sigma^2 I_{n \times n}$$

(Homo & Non auto)

$n \times n$ $n \times n$

$$\hat{\beta}_{GLS} = (X' \hat{\Sigma}^{-1} X)^{-1} X' \hat{\Sigma}^{-1} Y$$

$$\hat{\beta}_{OLS} = (X' (\hat{\sigma}^2 I)^{-1} X)^{-1} X' (\hat{\sigma}^2 I)^{-1} Y$$

$$= \hat{\sigma}^2 \cdot \frac{1}{\hat{\sigma}^2} (X' I^{-1} X)^{-1} X' I^{-1} Y$$

$$= (X' X)^{-1} X' Y$$

Heteroscedasticity

$$\hat{\Sigma} = \begin{bmatrix} \hat{\sigma}_1^2 & 0 & \dots & 0 \\ 0 & \hat{\sigma}_2^2 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & \hat{\sigma}_n^2 \end{bmatrix}$$

$$\hat{\beta}_{FGLS} = (X' \hat{\Sigma}^{-1} X)^{-1} X' \hat{\Sigma}^{-1} Y$$

Feasible GLS