

Regression

$$\hat{\beta} = (Z'Z)^{-1}Z'y, \quad \hat{y} = Z\hat{\beta} = Z(Z'Z)^{-1}Z'y = Hy$$

$$\hat{e} = y - \hat{y} = y - Z\hat{\beta} = (I - H)y$$

→ symmetric
→ Idempotent

$$Z'\hat{e} = 0, \quad \hat{y}'\hat{e} = 0$$

⇒ Sum of squares decomposition:

$$\hat{y}'\hat{e} = 0$$

$$TSS: y'y = \sum_{j=1}^n y_j^2 = y_1^2 + y_2^2 + y_3^2 + \dots + y_n^2$$

$$= (\hat{y} + y - \hat{y})'(\hat{y} + y - \hat{y})$$

$$= (\hat{y} + \hat{e})'(\hat{y} + \hat{e}) = \hat{y}'y + \hat{e}'\hat{e}$$

$$1'\hat{e} = 0 \quad \text{because mean of } e \text{ is zero}$$

$$\sum_{j=1}^n \hat{e}_j = \sum_{j=1}^n (y_j - \hat{y}_j) = \sum_{j=1}^n y_j - \sum_{j=1}^n \hat{y}_j = 0$$

$$= \boxed{\bar{y} = \bar{\hat{y}}} \rightarrow \text{averages are same.}$$

$$y'y - n\bar{y}^2 = \hat{y}'\hat{y} - n(\bar{\hat{y}})^2 + \hat{e}'\hat{e}$$

or

$$\sum_{j=1}^n (y_j - \bar{y})^2 = \sum_{j=1}^n (\hat{y}_j - \bar{\hat{y}})^2 + \sum_{j=1}^n \hat{e}_j^2$$

TSS about
mean

RSS

Regression sum of
sq

Residual sum of
sq

What proportion of the total variation is explained by the predictor variables.

Coefficient of determination

$$R^2 = 1 - \frac{\text{Residual SS}}{\text{TSS}}$$



Used to check
whether model
is good fit
or not.

$$0 \leq R^2 \leq 1$$

$R^2 \approx 1$ good fit

$R^2 \approx 0$ Not good fit

$$= 1 - \frac{\sum_{j=1}^n \hat{e}_j^2}{\sum_{j=1}^n (y_j - \bar{y})^2}$$

$$= \frac{\sum (\hat{y}_j - \bar{\hat{y}})^2}{\sum (y_j - \bar{y})^2}$$

Sampling Properties $\hat{\beta}, \hat{\epsilon}$

$$\begin{aligned}
 E(\hat{\beta}) &= E[(Z'Z)^{-1}Z'Y] \\
 &= E[(Z'Z)^{-1}Z'(Z\beta + \epsilon)] \\
 &= \underbrace{(Z'Z)^{-1}Z'Z}_{I} \beta + E(\epsilon) \\
 &= \beta \quad \text{where } E(\epsilon) = 0
 \end{aligned}$$

OLS $\hat{\beta}$ is unbiased.

$$\text{Cov}(\hat{\beta}) = \sigma^2 (Z'Z)^{-1}$$

$$\text{Cov}(\hat{\beta}) = E[(\hat{\beta} - \beta)(\hat{\beta} - \beta)']$$

$$\hat{\beta} = (Z'Z)^{-1}Z'(Z\beta + \epsilon) \quad \text{where } E[(Z'Z)^{-1}Z'\epsilon] = 0$$

$$= I\beta + (Z'Z)^{-1}Z'\epsilon$$

$$\hat{\beta} - \beta = (Z'Z)^{-1}Z'\epsilon$$

$$\text{Cov}(\hat{\beta}) = E[(Z'Z)^{-1}Z'\epsilon \epsilon' Z (Z'Z)^{-1}]$$

$$= (Z'Z)^{-1}Z' E(\epsilon \epsilon') Z (Z'Z)^{-1}$$

$$\begin{aligned}
 \text{Cov}(\epsilon) &= \sigma^2 I \\
 E[(\epsilon - 0)(\epsilon - 0)'] &= \sigma^2 I
 \end{aligned}$$

$$= \sigma^2 \mathbf{I} (\mathbf{z}'\mathbf{z})^{-1} \mathbf{z}'\mathbf{z} (\mathbf{z}'\mathbf{z})^{-1} = \sigma^2 (\mathbf{z}'\mathbf{z})^{-1}$$

$$* \quad E(\hat{\beta}) = \beta, \quad \text{Cov}(\hat{\beta}) = \sigma^2 (\mathbf{z}'\mathbf{z})^{-1}$$

$$E(\hat{e}), \quad \text{Cov}(\hat{e}) = \sigma^2 (\mathbf{I} - \mathbf{H})$$

$$\hat{e} = (\mathbf{I} - \mathbf{H})\mathbf{y} = (\mathbf{I} - \mathbf{H})(\mathbf{z}\beta + \mathbf{e})$$

$$= (\mathbf{I} - \mathbf{z}(\mathbf{z}'\mathbf{z})^{-1}\mathbf{z}')\beta + (\mathbf{I} - \mathbf{H})\mathbf{e}$$

$$= \mathbf{0} + (\mathbf{I} - \mathbf{H})\mathbf{e}$$

$$E(\hat{e}) = \mathbf{0} \quad \checkmark$$

$$\text{Cov}(\hat{e}) = \text{Cov}(\mathbf{I} - \mathbf{H})\mathbf{y}$$

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

* $\hat{\beta}$ and \hat{e} are uncorrelated

Gauss least square theorem

OLS is BLUE

Best linear unbiased estimator

$$\hat{\beta} = \mathbf{C}\mathbf{y}$$

$$\hat{\beta}_0 = c_1 y_1 + c_2 y_2 + \dots + c_n y_n$$

Variance of $\hat{\beta}$ is minimum among all other linear estimators.