

第一部份：

基本假設與符號

設 $Y = X\beta + \varepsilon$ ，其中 $X: n \times p, \text{rank}(X) = p$,
 $\beta = \begin{pmatrix} \beta_0 \\ \beta_1 \\ \vdots \\ \beta_{p-1} \end{pmatrix} : p \times 1$,

$\varepsilon \sim N_n(0, \sigma^2 I_n)$.

令 $\hat{\beta} = \begin{pmatrix} \hat{\beta}_0 \\ \hat{\beta}_1 \\ \vdots \\ \hat{\beta}_{p-1} \end{pmatrix}$ 是 β 的 LSE, $\hat{Y} = \begin{pmatrix} \hat{Y}_1 \\ \vdots \\ \hat{Y}_n \end{pmatrix} = X\hat{\beta}$,

$e = \begin{pmatrix} e_1 \\ \vdots \\ e_n \end{pmatrix} = Y - \hat{Y}$.

一. 令 $H = X(X'X)^{-1}X' = (h_{ij})$.

證明 (i) $\sum_{i=1}^n h_{ii} = p, \forall 1 \leq i \leq n, h_{ii} = h_{ii}^2 + \sum_{j \neq i} h_{ij}^2$

(12%) (ii) $\forall 1 \leq i \leq n, 0 \leq h_{ii} \leq 1$

(iii) 當 $h_{ii} = 0$ 時, $\hat{Y}_i = 0$

當 $h_{ii} = 1$ 時, $\hat{Y}_i = Y_i$

二. 若 A, D 對稱且必要的逆矩陣都存在,

$$\text{則 } \begin{pmatrix} A & B \\ B' & D \end{pmatrix}^{-1} = \begin{pmatrix} \bar{A}^{-1} + F E^{-1} F' & -F E^{-1} \\ -E^{-1} F' & E^{-1} \end{pmatrix}$$

其中 $E = D - B'A^{-1}B, F = \bar{A}^{-1}B$. (不必證明)

利用上述公式做下列的問題.

科目：迴歸分析專題

P2

- 設 $X = (x_0, x_1, \dots, x_{p-1}) = (W, x_{p-1})$
- 證明 (i) $|X'X| = |W'W| (x_{p-1}' x_{p-1} - x_{p-1}' W (W'W)^{-1} W' x_{p-1})$
- (18%) (ii) $\text{Var}[\hat{\beta}_{p-1}] \geq \sigma^2 (x_{p-1}' x_{p-1})^{-1}$
- (iii) 在 (ii) 中等號成立的充要條件為
 $\forall i = 0, 1, 2, \dots, p-2, x_i' x_{p-1} = 0$

三. 設 $A: q \times p$, 且 $\text{rank}(A) = q$. 在 $A\beta = c$ 的限制 (10%) 下, 求 β 的 LSE.

四. 設
$$\begin{pmatrix} Y_1 \\ \vdots \\ Y_n \end{pmatrix} = \begin{pmatrix} x_1' \\ \vdots \\ x_n' \end{pmatrix} \beta + \varepsilon$$

令 $S^2 = \frac{\sum_{i=1}^n e_i^2}{n-p}$ = 殘差均方 (residual mean square)

$S_{(i)}^2 =$ 去掉第 i 個資料點 (Y_i, x_i') 後所得之殘差均方.

若 $h_{ii} \neq 1$, 求 $\frac{e_i}{S_{(i)} \sqrt{1-h_{ii}}}$ 的機率分配,

(10%) 其中 h_{ii} 的定義如第一題中.

科目：迴歸分析專題

第二部份

迴歸分析

一. For the model $y = X\beta + \varepsilon$, with full rank of X , and $\varepsilon \sim N(0, \sigma^2 I_n)$.

Let $\hat{\beta} = (X'X)^{-1}X'y$, $S^2 = [y'y - \hat{y}'\hat{y}]/n-2 = SSE/n-2$, where $\hat{y} = X\hat{\beta}$.

a) Show that $\hat{\beta}$ and S^2 are joint sufficient statistics for β and σ^2 .

b) Show that $\hat{\beta}$ and S^2 are independent.

c) Show that $\hat{\beta}$ and S^2 are efficient.

二. For the model $y = X\beta + \varepsilon$, where X may not have full rank, let $\hat{\beta}$ be any solution of the normal equations $(X'X)\beta = X'y$.

a) Show that $\hat{\beta}$ minimizes SSE.

b) If $W\beta$ is estimable, then the BLUE of $W\beta$ is $W\hat{\beta}$.

三. For the model $y = X\beta + \varepsilon$ with $E(\varepsilon) = 0$, $\text{var}(\varepsilon) = \sigma^2 I_n$, but with no intercept term, let $\hat{\beta}$ be the estimator obtained by the least squares method.

a) Show that the sum of the elements of the estimated residual vector is not equal to zero. And prove that $X'\hat{\varepsilon} = 0$.

b) Show that $R^2 = 1 - \frac{\hat{\varepsilon}'\hat{\varepsilon}}{(y-\bar{y})(y-\bar{y})}$ can become negative. For which model this R^2 will be assumed to satisfy $0 \leq R^2 \leq 1$.