

(Previous class not attended)

DFBETA<sub>j(i)</sub>

↑ difference in Betas  
↑  $\hat{\beta}_j$  when the  $i^{th}$  observation

$(y_i, x_i^T)$  is not present in the data and the estimate of  $\hat{\beta}_j$  when the whole data is used.

Belsey, Kline, Welsch (1980)

$\hat{\beta}_j - \hat{\beta}_{j(i)} \sim ?$

$$\underline{L}^T (\hat{\beta} - \hat{\beta}_{(i)}) = \hat{\beta}_j - \hat{\beta}_{j(i)}$$

$$\underline{L}^T = (0 \ 0 \ 0 \ \dots \ 1 \ 0 \ \dots \ 0)$$

$$\hat{\beta} - \hat{\beta}_{(i)} = \frac{(X^T X)^{-1} x_i e_i}{1 - h_{ii}} \quad \text{--- (1)}$$

Denote  $R = (X^T X)^{-1} X^T$

$$\Rightarrow R R^T = (X^T X)^{-1} X^T X = (X^T X)^{-1} = (C_{ij})$$

$$\hat{\beta} \sim N(\beta, \sigma^2 C)$$

$$\underline{L}^T (\hat{\beta} - \hat{\beta}_{(i)}) = \frac{r_{ji} e_i}{(1 - h_{ii})} = \hat{\beta}_j - \hat{\beta}_{j(i)}$$

$r_{ji}$  is the  $j, i^{th}$  element of  $R$ .

$$\sqrt{L^T (\hat{\beta} - \hat{\beta}_{(i)})} = (\sigma^2 r_{ji}^T r_{ji}) (1 - h_{ii}) = \sigma^2 c_{jj} (1 - h_{ii}) \quad \text{(check)}$$

$$\frac{L^T (\hat{\beta} - \hat{\beta}_{(i)})}{\sqrt{\sigma^2 c_{jj} (1 - h_{ii})}} = \frac{\hat{\beta}_j - \hat{\beta}_{j(i)}}{\sqrt{\sigma^2 c_{jj} (1 - h_{ii})}}$$

If we use the estimated value of  $\sigma^2$  and  $S_{ii}^2$  then the estimate will be

$$\frac{\hat{\beta}_j - \hat{\beta}_{j(i)}}{\sqrt{S_{ii}^2 r_{ji}^T r_{ji}}} = \frac{\hat{\beta}_j - \hat{\beta}_{j(i)}}{\sqrt{c_{jj} S_{ii}^2}}$$

$$\frac{\hat{\beta}_j - \hat{\beta}_{j(i)}}{\sqrt{r_{ji}^T r_{ji} (S_{ii}^2)}} = \left( \frac{r_{ji} e_i}{(1 - h_{ii})} \right) / \sqrt{r_{ji}^T r_{ji} S_{ii}^2}$$

$$\frac{r_{ji}}{\sqrt{r_{ji}^T r_{ji} S_{ii}^2}} \frac{e_i}{\sqrt{1 - h_{ii}}} \frac{1}{\sqrt{1 - h_{ii}}}$$

$$= \left( \frac{r_{ji}}{\sqrt{r_{ji}^T r_{ji}}} \right) \left( \frac{e_i}{\sqrt{S_{ii}^2 (1 - h_{ii})}} \right) \left( \frac{1}{\sqrt{1 - h_{ii}}} \right)$$

t-statistic

$$= \frac{r_{ji}}{\sqrt{r_{ji}^T r_{ji}}} t_i^o \frac{1}{\sqrt{1 - h_{ii}}}$$

We can consider there is an influence point if  $|DF\beta_{eta,j(i)}| > \frac{2}{\sqrt{n}}$

$$DF\text{Fits}_0 = \frac{y_i - \hat{y}_{(i)}}{\sqrt{S_{ii}^2 h_{ii}}} = \frac{x_i^T (\hat{\beta} - \hat{\beta}_{(i)})}{\sqrt{S_{ii}^2 h_{ii}}}$$

$$= \frac{x_i^T (X^T X)^{-1} x_i e_i}{(1 - h_{ii}) \sqrt{S_{ii}^2 h_{ii}}}$$

$$= \frac{h_{ii}}{(1 - h_{ii})} \frac{e_i}{\sqrt{S_{ii}^2 h_{ii}}} = \left( \frac{h_{ii}}{(1 - h_{ii})} \right) \left( \frac{e_i}{\sqrt{S_{ii}^2 (1 - h_{ii})}} \right)$$

$$= \left( \frac{h_{ii}}{(1 - h_{ii})} \right)^{1/2} t_i^o$$

$|DF\text{Fits}| > \frac{2(K+1)}{n}$  then it is an influence point

General Variance of  $\hat{\beta}$  and  $\hat{\beta}$ .

$$\begin{aligned} \text{Cor Ratio} &= \left| \frac{\text{Var}(\hat{\beta}_{(i)})}{\text{Var}(\hat{\beta})} \right| \\ &= \left| \frac{\frac{1}{\sigma^2} (X_{(i)}^T X_{(i)})^{-1}}{\frac{1}{\sigma^2} (X^T X)^{-1}} \right| \\ &= \left| \frac{S_{ii}^2 (X_{(i)}^T X_{(i)})^{-1}}{MSR_{00} (X^T X)^{-1}} \right| \\ &= \left[ \frac{S_{ii}^2}{MSR_{00}} \right]^{K+1} \frac{(X_{(i)}^T X_{(i)})^{-1}}{(X^T X)^{-1}} \end{aligned}$$

$$|A + b e^T| = |A| (1 + e^T A^{-1} b)$$

$$\Rightarrow \frac{X^T X}{|X^T X| (1 - x_i^T (X^T X)^{-1} x_i)} = \frac{1}{1 - h_{ii}}$$

$$\therefore \left( \frac{S_{ii}^2}{MSR_{00}} \right)^{K+1} \frac{1}{1 - h_{ii}}$$