

# Prediction Intervals

Section 8.6, esp pp. 213-215

11.1

Denote a row of the  $X$  matrix by the column vector  $\tilde{x}_i$ . Then the regression model could be written  $Y_i = \tilde{x}_i' \beta + \epsilon_i$ .

Suppose we have a new vector of IV observations. Book calls it  $\tilde{x}_0$ . I would call it  $x_{n+1}$ . Then

$Y_0 = \tilde{x}_0' \beta + \epsilon_0$ ,  $\epsilon_0 \sim N(0, \sigma^2)$  & independent of  $\epsilon_1, \dots, \epsilon_n$

$E(Y_0) = \tilde{x}_0' \beta$ , and we can easily get a CONFIDENCE INTERVAL

$$\tilde{x}_0' \hat{\beta} - \tilde{x}_0' \beta \sim N(0, \sigma^2 \tilde{x}_0' (X'X)^{-1} \tilde{x}_0)$$

$$Z = \frac{\tilde{x}_0' \hat{\beta} - \tilde{x}_0' \beta}{\sqrt{\sigma^2 \tilde{x}_0' (X'X)^{-1} \tilde{x}_0}}, \quad W = \frac{SSE}{\sigma^2} \quad \text{ind}$$

$$T = \frac{Z}{\sqrt{W/(n-k-1)}} = \frac{\tilde{x}_0' \hat{\beta} - \tilde{x}_0' \beta}{\sqrt{\frac{\sigma^2 \tilde{x}_0' (X'X)^{-1} \tilde{x}_0 \cdot SSE}{\sigma^2 (n-k-1)}}$$

$$T = \frac{x_0' \hat{\beta} - x_0 \beta}{\sqrt{x_0' (X'X)^{-1} x_0 \frac{SSE}{n-k-1}}}$$

11.2

$$= \frac{x_0' \hat{\beta} - x_0 \beta}{\sqrt{x_0' (X'X)^{-1} x_0}}$$

$$1 - \alpha = P(-t_{\alpha/2} < T < t_{\alpha/2}) \text{ etc.}$$

Now suppose we want to PREDICT  
 $Y_{n+1}$  (keep calling it  $Y_0$ )

Best guess is surely  $x_0' \hat{\beta}$ , but  
 we want an INTERVAL around it,  
 a margin of error

CI captures  $E(Y_0)$  with prob  $1 - \alpha$

We want to capture  $Y_0$  - It's a  
 different problem, though related

$$Y_0 \sim N(x_0' \beta, \sigma^2)$$

and

$$x_0' \hat{\beta} \sim N(x_0' \beta, \sigma^2 x_0' (X'X)^{-1} x_0)$$

independent,  $S_0$

11.3

$$Y_0 - x_0' \hat{\beta} \sim N(0, \sigma^2 + \sigma^2 x_0' (X'X)^{-1} x_0)$$

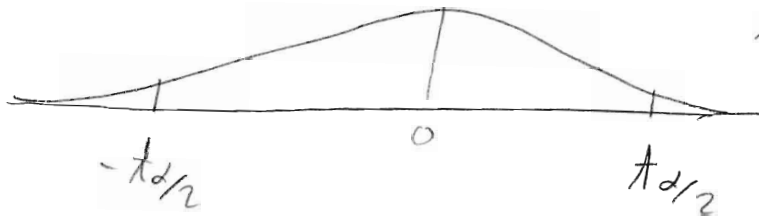
AND

$$Z = \frac{Y_0 - x_0' \hat{\beta}}{\sqrt{\sigma^2 (1 + x_0' (X'X)^{-1} x_0)}} \sim N(0, 1)$$

In dependent of  $\sigma^2$   $W = \frac{SSE}{\sigma^2}, S_0$

$$T = \frac{Z}{\sqrt{W/(n-k-1)}} = \frac{Y_0 - x_0' \hat{\beta}}{\sqrt{\sigma^2 (1 + x_0' (X'X)^{-1} x_0)} \sqrt{\frac{SSE}{\sigma^2} / (n-k-1)}}$$

$$= \frac{Y_0 - x_0' \hat{\beta}}{\sqrt{1 + x_0' (X'X)^{-1} x_0}} \sim t(n-k-1)$$



$t$  dist  
with  $n-k-1$  df

11.4

$$1 - \alpha = P(-t_{\alpha/2} < T < t_{\alpha/2})$$

$$= P(-t_{\alpha/2} < \frac{y_0 - x_0' \hat{\beta}}{\sigma \sqrt{1 + x_0' (X'X)^{-1} x_0}} < t_{\alpha/2})$$

$$= P(-t_{\alpha/2} \sigma \sqrt{\quad} < y_0 - x_0' \hat{\beta} < t_{\alpha/2} \sigma \sqrt{\quad})$$

$$= P(x_0' \hat{\beta} - t_{\alpha/2} \sigma \sqrt{\quad} < y_0 < x_0' \hat{\beta} + t_{\alpha/2} \sigma \sqrt{\quad}) \quad \text{or}$$

$$x_0' \hat{\beta} \pm t_{\alpha/2} \sigma \sqrt{1 + x_0' (X'X)^{-1} x_0}$$