

## Chapter 11

- Sums of squares:

Total sum of squares

$$SST = \sum_{i=1}^n (Y_i - \bar{Y})^2,$$

regression sum of squares

$$SSR = \sum_{i=1}^n (\hat{Y}_i - \bar{Y})^2,$$

error sum of squares

$$SSE = \sum_{i=1}^n (Y_i - \hat{Y}_i)^2.$$

Properties: a)  $SSR$  and  $SSE$  are independent, b)  $SST = SSR + SSE$ .

- Coefficient of determination (a measure of quality of fit)

$$R^2 = \frac{SSR}{SST} = 1 - \frac{SSE}{SST}.$$

$0 \leq R^2 \leq 1$ . The greater  $R^2$  the better fit.

- ANOVA approach. Testing

$$H_0 : \beta_1 = 0 \quad H_1 : \beta_1 \neq 0.$$

Test statistic

$$F = \frac{SSR}{SSE/(n-2)}.$$

Under  $H_0$ ,  $F$  has  $F$ -distribution with 1 and  $n-2$  degrees of freedom. Test:  $H_0$  is rejected if  $F \geq f_{\alpha,1,n-2}$ .

- Prediction. Observe  $Y_1, Y_2, \dots, Y_n$ . Predict  $Y_0 = \beta_0 + \beta_1 x_0 + \epsilon_0$ .

$(1-\alpha)$  confidence interval for  $EY_0$ :

$$\left[ \hat{Y}_0 - t_{\alpha/2, n-2} S \sqrt{\frac{1}{n} + \frac{(x_0 - \bar{x})^2}{S_{xx}}}, \hat{Y}_0 + t_{\alpha/2, n-2} S \sqrt{\frac{1}{n} + \frac{(x_0 - \bar{x})^2}{S_{xx}}} \right]$$

where  $\hat{Y}_0 = \hat{\beta}_0 + \hat{\beta}_1 x_0$ .

- $(1-\alpha)$  prediction interval for  $Y_0$ :

$$\left[ \hat{Y}_0 - t_{\alpha/2, n-2} S \sqrt{1 + \frac{1}{n} + \frac{(x_0 - \bar{x})^2}{S_{xx}}}, \hat{Y}_0 + t_{\alpha/2, n-2} S \sqrt{1 + \frac{1}{n} + \frac{(x_0 - \bar{x})^2}{S_{xx}}} \right].$$