

## Chapter 11

- Unbiased estimator of  $\sigma^2$  is

$$S^2 = \frac{1}{n-2} \sum_{i=1}^n (Y_i - \hat{\beta}_0 - \hat{\beta}_1 x_i)^2.$$

$S^2$  is independent on  $\bar{Y}$ ,  $\hat{\beta}_0$  and  $\hat{\beta}_1$ . Distribution of  $(n-2)S^2/\sigma^2$  is  $\chi^2$ -distribution with  $n-2$  degrees of freedom.

- Inference on the slope and intercept.

$(1-\alpha)$  confidence interval for  $\beta_1$ :

$$\left[ \hat{\beta}_1 - t_{\alpha/2,n-2} \frac{S}{\sqrt{\sum(x_i - \bar{x})^2}}, \hat{\beta}_1 + t_{\alpha/2,n-2} \frac{S}{\sqrt{\sum(x_i - \bar{x})^2}} \right].$$

$(1-\alpha)$  confidence interval for  $\beta_0$ :

$$\left[ \hat{\beta}_0 - t_{\alpha/2,n-2} \frac{S\sqrt{\sum x_i^2}}{\sqrt{n \sum(x_i - \bar{x})^2}}, \hat{\beta}_0 + t_{\alpha/2,n-2} \frac{S\sqrt{\sum x_i^2}}{\sqrt{n \sum(x_i - \bar{x})^2}} \right].$$

Testing  $H_0 : \beta_1 = \beta_{10}$ . Alternatives a)  $H_1 : \beta_1 > \beta_{10}$ , b)  $H_1 : \beta_1 < \beta_{10}$ , c)  $H_1 : \beta_1 \neq \beta_{10}$ . Significance level  $\alpha$ .

Test statistic

$$T = \frac{\hat{\beta}_1 - \beta_{10}}{S/\sqrt{S_{xx}}}$$

where  $S_{xx} = \sum(x_i - \bar{x})^2$ . Under  $H_0$ ,  $T$  has  $t$ -distribution with  $n-2$  degrees of freedom. Critical region: a)  $T \geq t_{\alpha/2,n-2}$ , b)  $T \leq -t_{\alpha/2,n-2}$ , c)  $|T| \geq t_{\alpha/2,n-2}$ .

Testing  $H_0 : \beta_0 = \beta_{00}$ . Alternatives a)  $H_1 : \beta_0 > \beta_{00}$ , b)  $H_1 : \beta_0 < \beta_{00}$ , c)  $H_1 : \beta_0 \neq \beta_{00}$ . Significance level  $\alpha$ .

Test statistic

$$T = \frac{\hat{\beta}_0 - \beta_{00}}{S\sqrt{\sum x_i^2/nS_{xx}}}$$

where  $S_{xx} = \sum(x_i - \bar{x})^2$ . Under  $H_0$ ,  $T$  has  $t$ -distribution with  $n-2$  degrees of freedom. Critical region: a)  $T \geq t_{\alpha/2,n-2}$ , b)  $T \leq -t_{\alpha/2,n-2}$ , c)  $|T| \geq t_{\alpha/2,n-2}$ .

- 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  $\hat{Y}_0 = \hat{\beta}_0 + \hat{\beta}_1 x_0 + \epsilon_0$ .  
 $(1-\alpha)$  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  $\hat{Y}_0 = \hat{\beta}_0 + \hat{\beta}_1 x_0$ .