

Note

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$$\hat{\beta}_0 = \bar{y} - \hat{\beta}_1 \bar{x}, \hat{\beta}_1 = \frac{S_{xy}}{S_{xx}}.$$

$$\text{Residual Standard Error } S = \sqrt{\frac{1}{n-2} \sum_{i=1}^n \hat{e}_i^2} = \sqrt{\frac{1}{n-2} \sum_{i=1}^n [y_i - \hat{y}_i]^2} = \sqrt{\frac{1}{n-2} \sum_{i=1}^n [y_i - (\hat{\beta}_0 + \hat{\beta}_1 * x_i)]^2}.$$

$$se(\hat{\beta}_1) = \frac{S}{\sqrt{S_{xx}}}, se(\hat{\beta}_0) = S \sqrt{\left(\frac{1}{n} + \frac{\bar{x}^2}{S_{xx}}\right)}.$$

1. Confidence Interval for β_1 : $\hat{\beta}_1 \pm t_{\alpha/2, n-2} * se(\hat{\beta}_1)$.

2. Confidence Interval for β_0 : $\hat{\beta}_0 \pm t_{\alpha/2, n-2} * se(\hat{\beta}_0)$.

1. Confidence Interval for mean Y (regression line) at $X = x^*$: $\hat{y}^* \pm t_{\alpha/2, n-2} * S \sqrt{\left(\frac{1}{n} + \frac{(x^* - \bar{x})^2}{S_{xx}}\right)}$

2. Prediction Interval for Single Y at $X = x^*$: $\hat{y}^* \pm t_{\alpha/2, n-2} * S \sqrt{\left(1 + \frac{1}{n} + \frac{(x^* - \bar{x})^2}{S_{xx}}\right)}$

Note that One sample t -test for mean has statistic $T = \frac{\bar{x} - \mu_0}{s/\sqrt{n}}$, where $s = \sqrt{\frac{S_{xx}}{n-1}}$, and the confidence interval is $\bar{x} \pm t_{\alpha/2, n-1} * \frac{s}{\sqrt{n}}$.

$$\text{Variation: } \sum_{i=1}^n (Y_i - \bar{Y})^2 = \sum_{i=1}^n (Y_i - \hat{Y}_i)^2 + \sum_{i=1}^n (\hat{Y}_i - \bar{Y})^2.$$

$$\rightarrow SST = SSE(RSS) + SSR(SSreg).$$

Simple notations:

$$SST = \sum_{i=1}^n (y_i - \bar{y})^2 = Syy.$$

$$SSR = \hat{\beta}_1 Sxy.$$

$$SSE = \sum_{i=1}^n \hat{e}_i^2 = Syy - \hat{\beta}_1 Sxy.$$

ANOVA table:

| Source of Variation | Degree of Freedom | Sum of Squares | Mean Squares | F | p -value |
|---------------------|-------------------|----------------|--------------|---|-------------------------|
| Regression | 1 | SSR | MSR | $\frac{MSR}{MSE} = \frac{SSR/1}{SSE/(n-2)}$ | $P(F < F_{df_1, df_2})$ |
| Residual | $n - 2$ | SSE | MSE | | |
| Total | $n - 1$ | SST | | | |