

(Hoff p. 152ff.)

Let \mathbf{y} be the n -dimensional column vector $(y_1, \dots, y_n)^T$ and let \mathbf{X} be the $n \times p$ matrix whose i th row is $\mathbf{x}_i = \{x_{i,1}, x_{i,2}, \dots, x_{i,p}\}$. Then the normal regression model is

$$\{\mathbf{y}|\mathbf{X}, \beta, \sigma^2\} \sim \text{multivariate normal}(\mathbf{X}\beta, \sigma^2\mathbf{I}),$$

where \mathbf{I} is the $p \times p$ identity matrix and

$$\mathbf{X}\beta = \begin{pmatrix} \mathbf{x}_1 \\ \mathbf{x}_2 \\ \vdots \\ \mathbf{x}_n \end{pmatrix} \begin{pmatrix} \beta_1 \\ \beta_2 \\ \vdots \\ \beta_p \end{pmatrix} = \begin{pmatrix} \beta_1 x_{1,1} + \dots + \beta_p x_{1,p} \\ \vdots \\ \beta_1 x_{n,1} + \dots + \beta_p x_{n,p} \end{pmatrix} = \begin{pmatrix} E[Y_1|\beta, \mathbf{x}_1] \\ \vdots \\ E[Y_n|\beta, \mathbf{x}_n] \end{pmatrix}$$

We compute the ordinary least squares estimates

$$\hat{\beta}_{ols} = (\mathbf{X}^T \mathbf{X})^{-1} \mathbf{X}^T \mathbf{y}$$

and

$$\hat{\sigma}_{ols}^2 = \frac{SSR(\hat{\beta}_{ols})}{(n-p)} = \frac{\sum (y_i - \hat{\beta}_{ols}^T x_i)^2}{(n-p)}.$$