

Summary

8 Sept 2020

Model

$$y_i = \beta_0 + \beta_1 x_{i1} + \dots + \beta_p x_{ip} + e_i \quad \text{for } i = 1, \dots, n$$

Or,

$$\underline{Y} = \underline{X} \underline{\beta} + \underline{\mathcal{E}}$$

where \underline{Y} is a $n \times 1$ column vector;

\underline{X} is a $n \times p'$ matrix, called "Design Matrix";

$\underline{\beta}$ is a $p' \times 1$ column vector;

n is the number of observations

$p' (= p + 1)$ is the number of unknown parameters in the model.

Assumptions

$$\left. \begin{array}{ll} (1) & E(e_i) = 0 \\ (2) & \text{Var}(e_i) = \sigma^2 \\ (3) & \text{Cov}(e_i, e_j) = 0 \\ (4) & e_i \sim N \end{array} \right\} \quad (\text{for } i \neq j) \quad e_i \stackrel{iid}{\sim} N(0, \sigma^2)$$

Or,

$$\underline{\mathcal{E}} \sim MN(\underline{0}, \sigma^2 \underline{I})$$

$$\Rightarrow \underline{Y} \sim MN(\underline{X}\underline{\beta}, \sigma^2 \underline{I})$$

Method of least squares

$p = 1$

Find $\hat{\beta}_0, \hat{\beta}_1$ such that $\sum_{i=1}^n \hat{e}_i^2$ is minimized, $\sum_{i=1}^n \hat{e}_i^2 = \sum_{i=1}^n (y_i - \hat{y}_i)^2$, where

- \hat{e}_i – define: Residual
- $\hat{y}_i = \hat{\beta}_0 + \hat{\beta}_1 x_{i1}$ for $i = 1, \dots, n$ – define: Fitted value