



[https://en.wikipedia.org/wiki/Regression\\_analysis](https://en.wikipedia.org/wiki/Regression_analysis)

[https://en.wikipedia.org/wiki/Ordinary\\_least\\_squares](https://en.wikipedia.org/wiki/Ordinary_least_squares)

$$\sum_{j=1}^p X_{ij}\beta_j = y_i, \quad (i = 1, 2, \dots, n), \quad \text{with } n > p$$

$$X = \begin{bmatrix} X_{11} & X_{12} & \cdots & X_{1p} \\ X_{21} & X_{22} & \cdots & X_{2p} \\ \vdots & \vdots & \ddots & \vdots \\ X_{n1} & X_{n2} & \cdots & X_{np} \end{bmatrix}, \quad \beta = \begin{bmatrix} \beta_1 \\ \beta_2 \\ \vdots \\ \beta_p \end{bmatrix}, \quad y = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix}.$$

$$X\beta = y,$$

$$\hat{\beta} = \arg \min_{\beta} S(\beta), \quad S(\beta) = \sum_{i=1}^n \left| y_i - \sum_{j=1}^p X_{ij}\beta_j \right|^2 = \|y - X\beta\|^2.$$

$$\hat{\beta} = (X^T X)^{-1} X^T y.$$

coefficient of determination:

$$R^2 = 1 - \frac{SS_{\text{res}}}{SS_{\text{tot}}}$$

$$SS_{\text{res}} = \sum_i (y_i - f_i)^2 = \sum_i e_i^2$$

$$SS_{\text{tot}} = \sum_i (y_i - \bar{y})^2$$