

MULTIPLE LINEAR REGRESSION

A Short Theoretical Tutorial

@S. S. Roy , Date: 27 July 2026

1. What It Is

Multiple linear regression extends the straight-line model to **two or more predictor variables**. A single response Y is expressed as a linear combination of k predictors $X_1, X_2, \dots, X_k$ plus a random error:

$$Y = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \dots + \beta_k X_k + \varepsilon$$

β_0 is the intercept; each β_j is a **partial regression coefficient**, giving the change in Y for a one-unit change in X_j when all other predictors are held fixed. ε is the random error, assumed to have mean zero, constant variance, and independence — and normality when tests are required. “Linear” refers to linearity in the parameters β , not in the predictors.

2. Matrix Form

With n observations the model is written compactly using matrices:

$$\mathbf{Y} = \mathbf{X}\boldsymbol{\beta} + \boldsymbol{\varepsilon}$$

where $\mathbf{Y}$ is the $n \times 1$ vector of responses, $\mathbf{X}$ is the $n \times (k+1)$ design matrix (a leading column of 1s for the intercept, then one column per predictor), $\boldsymbol{\beta}$ is the $(k+1) \times 1$ vector of unknown parameters, and $\boldsymbol{\varepsilon}$ is the $n \times 1$ error vector.

3. Least-Squares Estimation

The parameters are estimated by least squares, minimising the sum of squared residuals. The solution is the **normal-equations** result in matrix form:

$$\mathbf{b} = (\mathbf{X}'\mathbf{X})^{-1}\mathbf{X}'\mathbf{Y}$$

This single formula generalises the straight-line estimates $b_1 = S_{XY}/S_{XX}$ and $b_0 = \bar{Y} - b_1\bar{X}$. It requires $\mathbf{X}'\mathbf{X}$ to be **nonsingular** (invertible); $\mathbf{X}'\mathbf{X}$ is always square and symmetric. Once $\mathbf{b}$ is obtained, the fitted values are $\hat{\mathbf{Y}} = \mathbf{X}\mathbf{b}$ and the residuals are $\mathbf{Y} - \hat{\mathbf{Y}}$.

4. Key Points

- Each coefficient measures the effect of its predictor **after adjusting for all the others** — unlike simple regression, where only one predictor is present.
- Degrees of freedom for the error are **$n - (k + 1)$** , since $k + 1$ parameters (the intercept and k slopes) are estimated from the data.
- Individual coefficients are tested with **t-tests**; the overall usefulness of the model is tested with an **F-test**.
- Strong correlation among predictors (**multicollinearity**) makes $\mathbf{X}'\mathbf{X}$ nearly singular and the estimates unstable.

In one line: multiple linear regression fits $\mathbf{Y} = \mathbf{X}\boldsymbol{\beta} + \boldsymbol{\varepsilon}$ by least squares, giving $\mathbf{b} = (\mathbf{X}'\mathbf{X})^{-1}\mathbf{X}'\mathbf{Y}$ — the same

principle as the straight line, expressed in matrix form for many predictors.