

Linear and Nonlinear Models

A statistical estimation problem is nonlinear if the estimating equations—the equations whose solution yields the parameter estimates—depend on the parameters in a nonlinear fashion. Such estimation problems typically have no closed-form solution and must be solved by iterative, numerical techniques.

Nonlinearity in the mean function is often used to distinguish between linear and nonlinear models. A model has a nonlinear mean function if the derivative of the mean function with respect to the parameters depends on at least one other parameter. Consider, for example, the following models that relate a response variable Y to a single regressor variable x :

$$E[Y|x] = \beta_0 + \beta_1 x$$

$$E[Y|x] = \beta_0 + \beta_1 x + \beta_2 x^2$$

$$E[Y|x] = \beta + x/\alpha$$

In these expressions, $E[Y|x]$ denotes the expected value of the response variable Y at the fixed value of x . (The conditioning on x simply indicates that the predictor variables are assumed to be non-random. Conditioning is often omitted for brevity in this and subsequent chapters.)

The first model in the previous list is a simple linear regression (SLR) model. It is linear in the parameters β_0 and β_1 since the model derivatives do not depend on unknowns:

$$\frac{\partial}{\partial \beta_0} (\beta_0 + \beta_1 x) = 1$$

$$\frac{\partial}{\partial \beta_1} (\beta_0 + \beta_1 x) = x$$

The model is also linear in its relationship with x (a straight line). The second model is also linear in the parameters, since

$$\frac{\partial}{\partial \beta_0} (\beta_0 + \beta_1 x + \beta_2 x^2) = 1$$

$$\frac{\partial}{\partial \beta_1} (\beta_0 + \beta_1 x + \beta_2 x^2) = x$$

$$\frac{\partial}{\partial \beta_2} (\beta_0 + \beta_1 x + \beta_2 x^2) = x^2$$

However, this second model is *curvilinear*, since it exhibits a curved relationship when plotted against x . The third model, finally, is a nonlinear model since

$$\frac{\partial}{\partial \beta}(\beta + x/\alpha) = 1$$

$$\frac{\partial}{\partial \alpha}(\beta + x/\alpha) = -\frac{x}{\alpha^2}$$