

The General Linear Model

In a **general linear model**

$$y_i = \beta_0 + \beta_1 x_{1i} + \dots + \beta_p x_{pi} + \epsilon_i$$

the **response** $y_i, i = 1, \dots, n$ is modelled by a linear function of **explanatory** variables $x_j, j = 1, \dots, p$ plus an error term.

General and Linear

Here **general** refers to the dependence on potentially more than one explanatory variable, v.s. the **simple linear model**:

$$y_i = \beta_0 + \beta_1 x_i + \epsilon_i$$

The model is *linear in the parameters*, e.g.

$$y_i = \beta_0 + \beta_1 x_1 + \beta_2 x_1^2 + \epsilon_i$$

$$y_i = \beta_0 + \gamma_1 \delta_1 x_1 + \exp(\beta_2) x_2 + \epsilon_i$$

but not e.g.

$$y_i = \beta_0 + \beta_1 x_1^{\beta_2} + \epsilon_i$$

$$y_i = \beta_0 \exp(\beta_1 x_1) + \epsilon_i$$

Error structure

We assume that the errors ϵ_i are independent and identically distributed such that

$$E[\epsilon_i] = 0$$

and $\text{var}[\epsilon_i] = \sigma^2$

Typically we assume

$$\epsilon_i \sim N(0, \sigma^2)$$

as a basis for inference, e.g. t-tests on parameters.

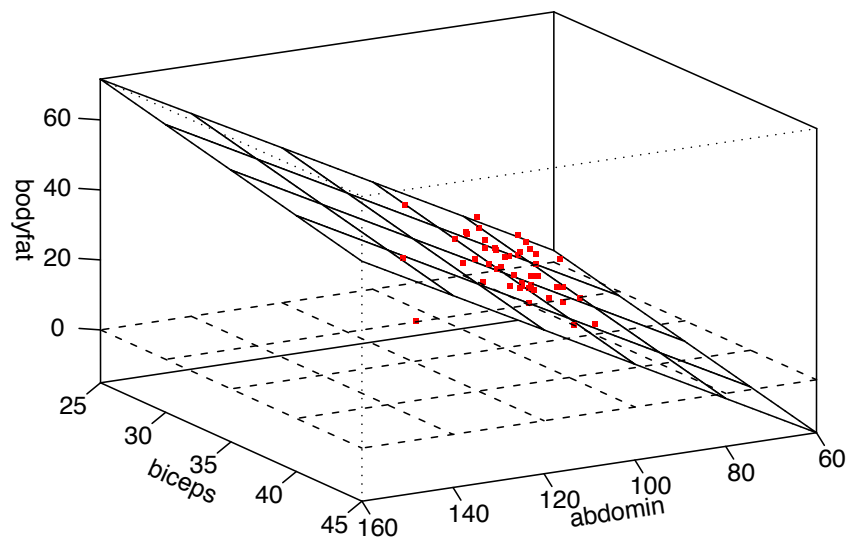
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Some Examples

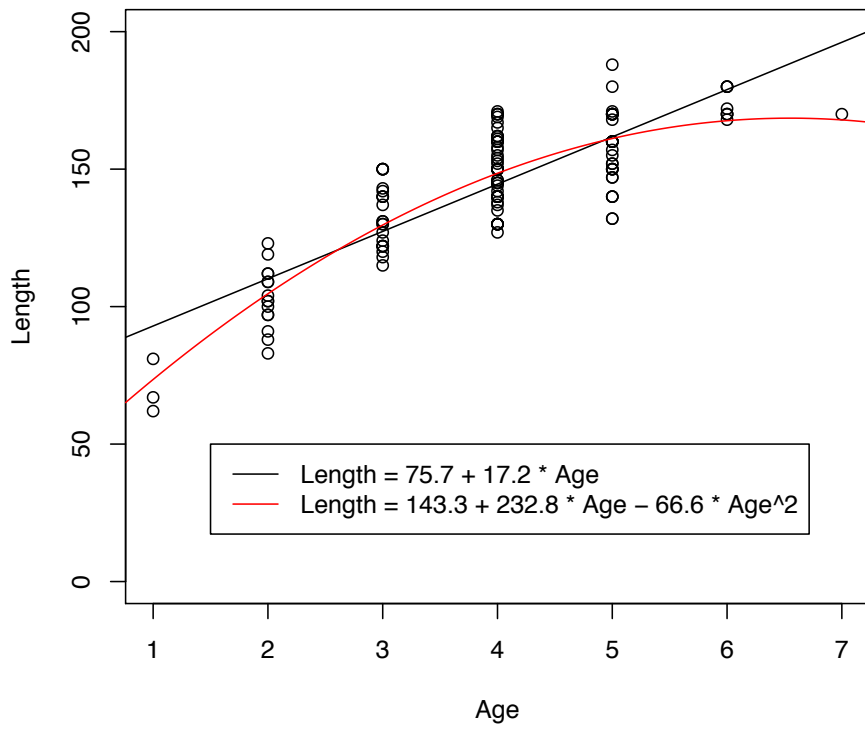
$$\text{bodyfat} = -14.59 + 0.7 * \text{biceps} - 0.9 * \text{abdomin}$$



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$$\text{particle size}_{ij} = \text{operator}_i + \text{resin}_j + \text{operator}:\text{resin}_{ij}$$

