



## Developing Equations for Computing Regression Coefficients

Let a dependent variable  $Y$  and an independent variable  $x$  be related by an estimated regression function

$$Y = a + bx \quad (\text{B.1})$$

Let  $Y_i$  be an estimate and  $y_i$  be an observed value of  $Y$  for a corresponding value  $x_i$  for  $x$ . Estimates of  $a$  and  $b$  can be obtained by minimizing the sum of the squares of the differences ( $R$ ) between  $Y_i$  and  $y_i$  for a set of observed values, where

$$R = \sum_{i=1}^n (y_i - Y_i)^2 \quad (\text{B.2})$$

Substituting  $(a + bx_i)$  for  $Y_i$  in EqB.2, we obtain

$$R = \sum_{i=1}^n (y_i - a - bx_i)^2 \quad (\text{B.3})$$

Differentiating  $R$  partially with respect to  $a$ , then with respect to  $b$ , and equating each to zero, we obtain

$$\frac{\partial R}{\partial a} = -2 \sum_{i=1}^n (y_i - a - bx_i) = 0 \quad (\text{B.4})$$

$$\frac{\partial R}{\partial b} = -2 \sum_{i=1}^n x_i (y_i - a - bx_i) = 0 \quad (\text{B.5})$$

From EqB.4, we obtain

$$\sum_{i=1}^n y_i = na + b \sum_{i=1}^n x_i \quad (\text{B.6})$$

giving

$$a = \frac{1}{n} \sum_{i=1}^n y_i - \frac{b}{n} \sum_{i=1}^n x_i \quad (\text{B.7})$$

From EqB.5, we obtain

$$\sum_{i=1}^n x_i y_i = a \sum_{i=1}^n x_i + b \sum_{i=1}^n x_i^2 \quad (\text{B.8})$$

Substituting for  $a$ , we obtain

$$b = \frac{\sum_{i=1}^n x_i y_i - \frac{1}{n} \left( \sum_{i=1}^n x_i \right) \left( \sum_{i=1}^n y_i \right)}{\sum_{i=1}^n x_i^2 - \frac{1}{n} \left( \sum_{i=1}^n x_i \right)^2} \quad (\text{B.9})$$

where

$n$  = number of sets of observation

$x_i$  =  $i$ th observation for  $x$

$y_i$  =  $i$ th observation for  $y$

Eqns B.8 and B.9 may be used to obtain estimated values for  $a$  and  $b$  in EqB.1. To test the suitability of the regression function obtained, the coefficient of determination,  $R^2$  (which indicates to what extent values of  $Y_i$  obtained from the regression function agree with observed values  $y_i$ ) is determined from the expression

$$R^2 = \frac{\sum_{i=1}^n (Y_i - \bar{y})^2}{\sum_{i=1}^n (y_i - \bar{y})^2} \quad (\text{B.10})$$

which is also written as

$$R^2 = \frac{\sum_{i=1}^n \left( x_i y_i - n \bar{x} \bar{y} \right)^2}{\sum_{i=1}^n \left( x_i^2 - n \bar{x}^2 \right) \left( \sum_{i=1}^n y_i^2 - n \bar{y}^2 \right)} \quad (\text{B.11})$$

The closer the value of  $R^2$  is to 1, the more suitable the estimated regression function for the data.