

to make these computations simpler, note that

$$\sum_{i=1}^n (X_{1i} - \bar{X}_1)^2 = \sum_{i=1}^n X_{1i}^2 - \frac{\left(\sum_{i=1}^n X_{1i}\right)^2}{n}$$

$$\sum_{i=1}^n (X_{2i} - \bar{X}_2)^2 = \sum_{i=1}^n X_{2i}^2 - \frac{\left(\sum_{i=1}^n X_{2i}\right)^2}{n}$$

$$\sum_{i=1}^n (X_{1i} - \bar{X}_1)(Y_i - \bar{Y}) = \sum_{i=1}^n X_{1i} Y_i - \frac{\left(\sum_{i=1}^n X_{1i}\right) \left(\sum_{i=1}^n Y_i\right)}{n}$$

$$\sum_{i=1}^n (X_{2i} - \bar{X}_2)(Y_i - \bar{Y}) = \sum_{i=1}^n X_{2i} Y_i - \frac{\left(\sum_{i=1}^n X_{2i}\right) \left(\sum_{i=1}^n Y_i\right)}{n}$$

$$\sum_{i=1}^n (X_{1i} - \bar{X}_1)(X_{2i} - \bar{X}_2) = \sum_{i=1}^n X_{1i} X_{2i} - \frac{\left(\sum_{i=1}^n X_{1i}\right) \left(\sum_{i=1}^n X_{2i}\right)}{n}$$

The following example illustrates how least-squares estimates of  $A$ ,  $B_1$ ,

normal) equations:

$$\sum_{i=1}^n Y_i = na + b_1 \sum_{i=1}^n X_{1i} + b_2 \sum_{i=1}^n X_{2i}$$

$$\sum_{i=1}^n X_{1i} Y_i = a \sum_{i=1}^n X_{1i} + b_1 \sum_{i=1}^n X_{1i}^2 + b_2 \sum_{i=1}^n X_{1i} X_{2i}$$

$$\sum_{i=1}^n X_{2i} Y_i = a \sum_{i=1}^n X_{2i} + b_1 \sum_{i=1}^n X_{1i} X_{2i} + b_2 \sum_{i=1}^n X_{2i}^2$$

(12.4)

Solving these equations for  $a$ ,  $b_1$ , and  $b_2$ , we obtain the following results:

$$b_1 = \frac{\sum_{i=1}^n (X_{2i} - \bar{X}_2)^2 \sum_{i=1}^n (X_{1i} - \bar{X}_1)(Y_i - \bar{Y}) - \sum_{i=1}^n (X_{1i} - \bar{X}_1)(X_{2i} - \bar{X}_2) \sum_{i=1}^n (X_{2i} - \bar{X}_2)(Y_i - \bar{Y})}{\sum_{i=1}^n (X_{1i} - \bar{X}_1)^2 \sum_{i=1}^n (X_{2i} - \bar{X}_2)^2 - \left[ \sum_{i=1}^n (X_{1i} - \bar{X}_1)(X_{2i} - \bar{X}_2) \right]^2}$$

$$b_2 = \frac{\sum_{i=1}^n (X_{1i} - \bar{X}_1)^2 \sum_{i=1}^n (X_{2i} - \bar{X}_2)(Y_i - \bar{Y}) - \sum_{i=1}^n (X_{1i} - \bar{X}_1)(X_{2i} - \bar{X}_2) \sum_{i=1}^n (X_{1i} - \bar{X}_1)(Y_i - \bar{Y})}{\sum_{i=1}^n (X_{1i} - \bar{X}_1)^2 \sum_{i=1}^n (X_{2i} - \bar{X}_2)^2 - \left[ \sum_{i=1}^n (X_{1i} - \bar{X}_1)(X_{2i} - \bar{X}_2) \right]^2}$$

$$a = \bar{Y} - b_1 \bar{X}_1 - b_2 \bar{X}_2$$

(12.5)