**Problem statement** Given *n* data points  $(x_1, y_1), \ldots, (x_n, y_n)$ , we may seek a linear function y = mx + b that best fits the data. The **linear least-squares fit** is the linear function f(x) = mx + b that minimizes the sum of the squares (see the Figure)  $E(m, b) = \sum_{n=1}^{n} (u_n - f(x_n))^2$ 

$$\sum_{\substack{nj=1\\ \text{Cl}}} (y_j - f(x_j))^2$$

Show that E is minimized for m and b satisfying

$$m\sum_{j=1}^{n} x_j + bn = \sum_{j=1}^{n} y_j$$
 and  $m\sum_{j=1}^{n} x_j^2 + b\sum_{j=1}^{n} x_j = \sum_{j=1}^{n} x_j y_j$ 

**Comment** This is problem 44 in the textbook's section 14.7. The result is quite important in practical computation. Several assertions must be verified: that E has one critical point which is a local minimum, and that this local minimum is actually an *absolute* minimum.

