## Additional Note on Linear Regression

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Suppose we have measured two variables X and Y (say height and weight). We consider drawing a regression line that characterizes their relationship. Since the line is expressed as Y = bX + a, our job is to determine coefficients b and a.

To achieve this goal we calculate the deviance of a line from the actual data. For each data point i, we can measure the difference or "error" between actual  $y_i$  and the "prediction"  $bx_i + a$  by

$$l_i = \{y_i - (bx_i + a)\}^2 \tag{1}$$

The square is required because we are interested just in the distance from the line to the data point, not a direction (whether the difference is positive or negative). If we have n data, we can sum up (1) to obtain

$$L = \sum_{i=1}^{n} \{y_i - (bx_i + a)\}^2$$
 (2)

The best line (i.e., a and b) minimizes this sum of squared errors. This is a quadratic function of a and b, so to minimize L we take its partial derivative with respect to a and b and set them to zero:

$$\frac{\partial L}{\partial b} = \sum y_i - b \sum x_i - na = 0$$

$$\frac{\partial L}{\partial b} = \sum x_i y_i - b \sum x_i^2 - a \sum x_i = 0$$

By solving this system of equation yields

$$b = \frac{\sum x_i y_i - n\bar{X}\bar{Y}}{\sum x_i^2 - n\bar{X}^2},$$
  
$$a = \bar{Y} - b\bar{X}.$$

Check by yourself that this b equals the regression coefficient Cov(X,Y)/Var(Y) introduced in the note. This procedure is called the method of least squares.