

# Prediction Intervals and Variation

Colby Community College

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## Formula

Given a fixed and known value  $x_0$ , the prediction interval for an individual  $y$  value is

$$\hat{y} - E < y < \hat{y} + E$$

where

$$E = t_{\alpha/2} s_e \sqrt{1 + \frac{1}{n} + \frac{n(x_0 - \bar{x})^2}{n(\sum x^2) - (\sum x)^2}} \text{ and } s_e = \sqrt{\frac{\sum (y - \hat{y})^2}{n - 2}}$$

and  $t_{\alpha/2}$  has  $n - 2$  degrees of freedom.

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## Note

We could narrow down the interval by using a much larger set of data.

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For the following definitions, we assume that we have a collection of paired data containing the sample point  $(x, y)$ , that  $\hat{y}$  is the predicted value of  $y$  (obtained by using the regression equation), and that the mean of the sample  $y$  values is  $\bar{y}$ .

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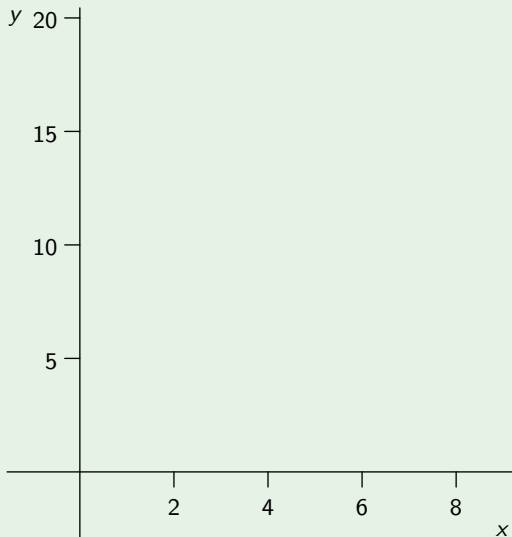
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The **unexplained deviation** is the vertical distance  $(y - \hat{y})$ , which is the vertical distance between the point  $(x, y)$  and the regression line.

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We assume the following:

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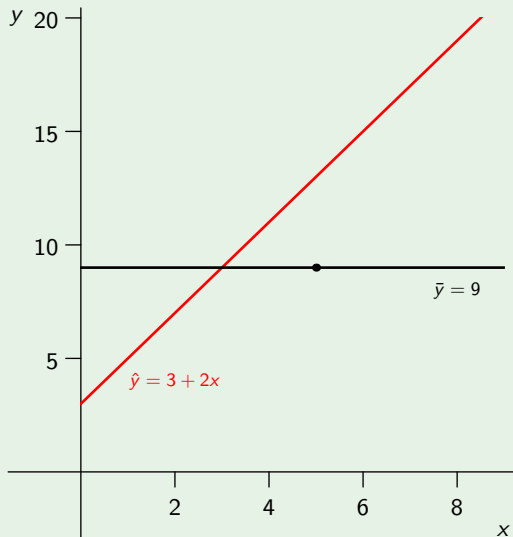
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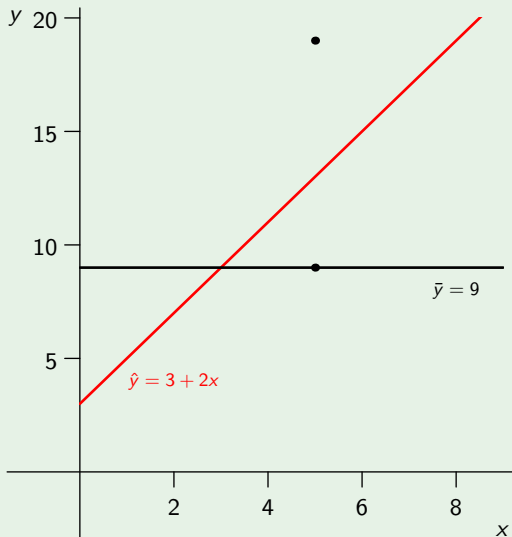


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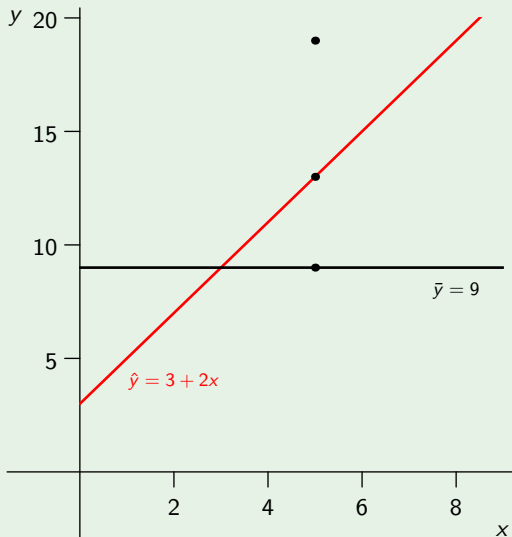
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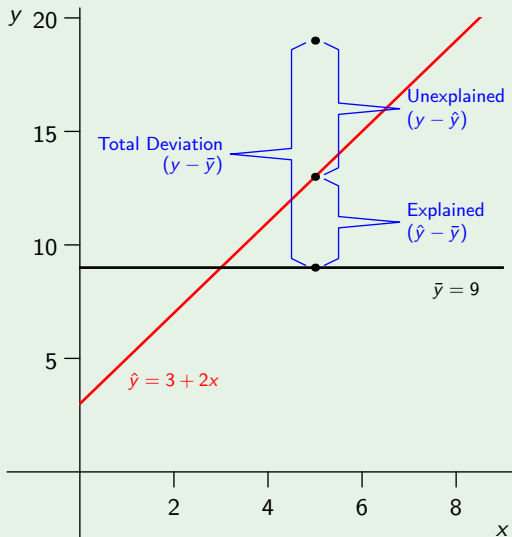
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We can either compute  $r^2$  using the variation or we can square the linear correlation coefficient  $r$ .

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### Note

The other 35.8% might be explained by some other factors. But it is pretty silly to seriously think that chocolate consumption in a country is going to directly effect the country's rate of Nobel Laureates.