

Standard Deviation and Covariance

Variance is the statistical second moment of a distribution.

The variance S_x^2 is defined as $\{ \sum (x_i - \bar{x})^2 \} / n$

where S_x is the standard deviation and $\bar{x} = (\sum x_i) / n$

$$\begin{aligned} \text{So } \{ \sum (x_i - \bar{x})^2 \} / n &= \{ \sum x_i^2 - \sum 2 x_i \bar{x} + \sum \bar{x}^2 \} / n \\ &= \{ (\sum x_i^2) / n - 2 \bar{x} (1/n) \sum x_i + \bar{x}^2 \} \end{aligned}$$

$$\text{but as } (1/n) \sum x_i = \bar{x} \quad S_x^2 = \{ (\sum x_i^2) / n - \bar{x}^2 \}$$

$$S_x = \sqrt{ \{ (\sum x_i^2) / n - \bar{x}^2 \} }$$

Covariance

In probability theory and statistics, **covariance** is a measure of the joint variability of two random variables. The sign of the covariance, therefore, shows the tendency in the linear relationship between the variables.

The covariance S_{xy} is defined as $\{ \sum (x_i - \bar{x})(y_i - \bar{y}) \} / n$

$$\begin{aligned} \text{Expanding } (x_i - \bar{x})(y_i - \bar{y}) \text{ gives } S_{xy} &= \{ \sum x_i y_i - \bar{x} \sum y_i - \bar{y} \sum x_i + \sum \bar{x} \bar{y} \} / n \\ &= \{ \sum x_i y_i \} / n - \bar{x} \bar{y} - \bar{x} \bar{y} + \bar{x} \bar{y} \\ &= \{ \sum x_i y_i \} / n - \bar{x} \bar{y} \end{aligned}$$

which again is pretty neat.

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