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## Variance of a discrete random variable

*Noun Phrase.* (Probability.)

The variance of a discrete [random variable](#) is a measure of the dispersion of the variable. The variance measures how different the values of the random variable tend to be from the [expectation](#), or mean, of the random variable. The more the values tend to differ from the expectation, the greater the variance.

The variance of a random variable is simply the variance of the [probability distribution](#) of the random variable. So, for discrete random variables, the variance is the probability-weighted mean of the squared deviations of the different values from their mean.

Let's look at two examples.

### Example 1.

Seventy-six percent of the students at a particular University of California campus are undergraduates, and the other 24% are graduates. If we choose a student at random from the campus and code the student as a "1" if the student is an undergraduate and as a "0" if the student is a graduate, then the student's standing is a random variable -- we'll call it  $X$  -- which can take on the value 0 or the value 1.

The probability distribution of  $X$  is as follows:

Value $x$ of $X$	0	1
$P(X = x)$	0.76	0.24

The expectation of  $X$  is  $E(X) = (0)(0.76) + (1)(0.24) = 0.24$ .

The variance of  $X$  is computed by

$$\begin{aligned}
 \text{Var}(X) &= (0 - E(X))^2 P(X=0) + (1 - E(X))^2 P(X=1) \\
 &= (0 - 0.24)^2 (0.76) + (1 - 0.24)^2 (0.24) \\
 &= 0.1824.
 \end{aligned}$$

### Example 2.

Let  $Y$  be the random variable which gives the number that appears when a fair die is rolled.

The probability distribution of  $Y$  is as follows:

Value $y$ of $Y$	1	2	3	4	5	6
$P(Y = y)$	1/6	1/6	1/6	1/6	1/6	1/6

The expectation of  $Y$  is

$$E(Y) = (1)\left(\frac{1}{6}\right) + (2)\left(\frac{1}{6}\right) + (3)\left(\frac{1}{6}\right) + (4)\left(\frac{1}{6}\right) + (5)\left(\frac{1}{6}\right) + (6)\left(\frac{1}{6}\right) = \frac{21}{6} = 3.5.$$

The variance of  $Y$  is then

$$\begin{aligned} \text{Var}(Y) &= (1 - 3.5)^2\left(\frac{1}{6}\right) + (2 - 3.5)^2\left(\frac{1}{6}\right) + (3 - 3.5)^2\left(\frac{1}{6}\right) + (4 - 3.5)^2\left(\frac{1}{6}\right) + (5 - 3.5)^2\left(\frac{1}{6}\right) + (6 - 3.5)^2\left(\frac{1}{6}\right) \\ &\approx 2.92. \end{aligned}$$

### Definition.

Let  $X$  be a discrete random variable that takes on the values  $x_1, x_2, \dots, x_n$  (and only those values). The *variance* of  $X$ , often denoted  $\text{Var}(X)$ , is defined to be

$$\text{Var}(X) = E[(X - E(X))^2] = \sum_{i=1}^n (x_i - E(X))^2 P(X = x_i),$$

where  $E(X)$  is the expectation of  $X$ .

The *standard deviation* of  $X$  is the positive square root of its variance ,

$$\sqrt{\text{Var}(X)}.$$

### Some Comments.

- The notion of variance of a discrete random variable agrees with the notion of [variance of a population](#). The difference between the two lies in the fact that the variance of a discrete random variable involves the [probabilities](#) of the values of the random variable, while the population variance involves the [relative frequencies](#) of the population values. Such a comparison may also be made between the notions of expectation of a random variable and [sample mean](#) or [population mean](#).

- The variance of the discrete random variable  $X$  may also be computed by the equivalent formula

$$\text{Var}(X) = E(X^2) - (E(X))^2.$$

This formula often leads to easier computation than the formula given in the definition. To see that the two formulas are equivalent, note that

$$\begin{aligned} \text{Var}(X) &= E[(X - E(X))^2] && \text{[by the definition of } \text{Var}(X)\text{]} \\ &= E[X^2 - 2XE(X) + (E(X))^2] \\ &= E(X^2) - 2E(X)E(X) + (E(X))^2 && \text{[by properties of the expectation } E(X)\text{]} \\ &= E(X^2) - (E(X))^2. \end{aligned}$$

### Properties of the variance.

- Let  $a$  be any [real number](#), and let  $X$  be a random variable. Then

$$\text{Var}(aX) = a^2 \text{Var}(X)$$

and

$$\text{Var}(X \pm a) = \text{Var}(X).$$

This second equation is a consequence of the fact that the deviations from the expectation do not change if you add (or subtract) a constant to all the values of the random variable.

- For any random variable  $X$ , we have  $\text{Var}(X) = 0$  if, and only if,  $X$  is constant.
- If  $X$  and  $Y$  are two [independent random variables](#), then

$$\text{Var}(X + Y) = \text{Var}(X) + \text{Var}(Y).$$