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## Expectation

*Noun.* (Probability.)

The *expectation* ( or *mean* , or *expected value*) of a [random variable](#) is the mean of the [probability distribution](#) of the random variable.

So, for a discrete random variable, the expectation is a [weighted mean](#) of the values that the variable may take on, with the weights being the probabilities.

Just as the mean of a probability distribution is a measure of central tendency for the distribution, the expectation of a random variable is a measure of the central tendency of the values that the random variable takes on.

The expectation of the random variable  $X$  is usually denoted by  $E(X)$ .

### Example 1:

Let  $D$  be the random variable giving the number obtained when a fair die is rolled. We specify the probability distribution of  $D$  in the table below, which contains the possible values of  $D$  and the probabilities that  $D$  takes on those values:

Value of $D$	1	2	3	4	5	6
Probability	$\frac{1}{6}$	$\frac{1}{6}$	$\frac{1}{6}$	$\frac{1}{6}$	$\frac{1}{6}$	$\frac{1}{6}$

So

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

Note a couple of things about this particular expectation. First, it is not a possible value of  $D$ . (It is impossible to obtain a 3.5 in a roll of a die.) This is not a problem, and in fact it motivates the idea of using the term *expectation* rather than *expected value*, since the expectation may not be a value that the random variable takes on. Second, the expectation in this case is equal to the [arithmetic mean](#) of the possible values of  $D$ , that is, the arithmetic mean of 0, 1, 2, 3, 4, 5, and 6. This is because  $D$  takes each of these values with equal probability.

**Example 2:**

Let  $S$  be the random variable giving the sum of the numbers when two, fair dice are rolled. The probability distribution of  $S$  is given in the table below, which contains the possible values of  $S$  and the probabilities that  $S$  takes on those values:

Value of $S$	2	3	4	5	6	7	8	9	10	11	12
Probability	$\frac{1}{36}$	$\frac{2}{36}$	$\frac{3}{36}$	$\frac{4}{36}$	$\frac{5}{36}$	$\frac{6}{36}$	$\frac{5}{36}$	$\frac{4}{36}$	$\frac{3}{36}$	$\frac{2}{36}$	$\frac{1}{36}$

So

$$E(S) = (2)\left(\frac{1}{36}\right) + (3)\left(\frac{2}{36}\right) + (4)\left(\frac{3}{36}\right) + (5)\left(\frac{4}{36}\right) + (6)\left(\frac{5}{36}\right) + (7)\left(\frac{6}{36}\right) \\ + (8)\left(\frac{5}{36}\right) + (9)\left(\frac{4}{36}\right) + (10)\left(\frac{3}{36}\right) + (11)\left(\frac{2}{36}\right) + (12)\left(\frac{1}{36}\right) = 7.$$

Note that the expectation of the sum of the numbers on the dice is equal to the sum of the expectations of the numbers on the individual dice, that is,  
 $7 = E(S) = E(D + D) = E(D) + E(D) = 3.5 + 3.5.$

We have the following general definition.

**Definition of expectation of a discrete random variable:**

Suppose that the probability distribution of the discrete random variable  $X$  is given by the following table:

Value of $X$	$x_1$	$x_2$	$x_3$	...	$x_j$	...	$x_k$
Probability	$p_1$	$p_2$	$p_3$	...	$p_j$	...	$p_k$

Then the *expectation* of  $X$  is given by

$$E(X) = x_1 p_1 + x_2 p_2 + x_3 p_3 + \dots + x_k p_k = \sum_{i=1}^k x_i p_i.$$

**Properties of the expectation:**

- If  $a$  is a constant and  $X$  a random variable, then

$$E(a) = a$$

and

$$E(aX) = aE(X).$$

- If  $X$  and  $Y$  are two random variables ([independent](#) or not), then

$$E(X + Y) = E(X) + E(Y).$$

As a special case, for any constant  $a$ ,

$$E(X + a) = E(X) + a.$$

- If  $X$  and  $Y$  are two [independent random variables](#), then

$$E(XY) = E(X)E(Y).$$

### See also:

- [Population mean](#)