



## 6.1. Binomial Distribution

We consider repeated and independent trials of an experiment with two outcomes; we call one of the outcomes *success* and the other outcome *failure*. Let  $p$  be the probability of success, so that  $q = 1 - p$  is the probability of failure. If we are interested in the number of successes and not in the order in which they occur, then the following theorem applies.

### Theorem

The probability of exactly  $k$  successes in  $n$  repeated trials is denoted and given by

$$b(k; n, p) = \binom{n}{k} p^k q^{n-k}$$

Here  $\binom{n}{k}$  is the binomial coefficient  $\left( \begin{matrix} n \\ k \end{matrix} \right)$ . Observe that the probability of no successes is  $q^n$ , and therefore the probability of at least one success is  $1 - q^n$ .

**Example 1:** A fair coin is tossed 6 times or, equivalently, six fair coins are tossed; call heads a success. Then  $n = 6$  and  $p = q = \frac{1}{2}$ .

(i) The probability that exactly two heads occur (i.e.  $k = 2$ ) is

$$b(2; 6, \frac{1}{2}) = \binom{6}{2} (\frac{1}{2})^2 (\frac{1}{2})^4 = \frac{15}{64}$$

(ii) The probability of getting at least four heads (i.e.  $k = 4, 5$  or  $6$ ) is

$$\begin{aligned} b(4; 6, \frac{1}{2}) + b(5; 6, \frac{1}{2}) + b(6; 6, \frac{1}{2}) &= \binom{6}{4} (\frac{1}{2})^4 (\frac{1}{2})^2 + \binom{6}{5} (\frac{1}{2})^5 (\frac{1}{2}) + \binom{6}{6} (\frac{1}{2})^6 \\ &= \frac{15}{64} + \frac{6}{64} + \frac{1}{64} = \frac{11}{32} \end{aligned}$$

(iii) The probability of no heads (i.e. all failures) is  $q^6 = (\frac{1}{2})^6 = \frac{1}{64}$ , and so the probability of at least one head is  $1 - q^6 = 1 - \frac{1}{64} = \frac{63}{64}$ .



**Example 2:** A fair die is tossed 7 times; call a toss a success if a 5 or a 6 appears. Then  $n = 7$ ,  $p = P(\{5, 6\}) = \frac{1}{3}$  and  $q = 1 - p = \frac{2}{3}$ .

(i) The probability that a 5 or a 6 occurs exactly 3 times (i.e.  $k = 3$ ) is

$$b(3; 7, \frac{1}{3}) = \binom{7}{3} (\frac{1}{3})^3 (\frac{2}{3})^4 = \frac{560}{2187}$$

(ii) The probability that a 5 or a 6 never occurs (i.e. all failures) is  $q^7 = (\frac{2}{3})^7 = \frac{128}{2187}$ ;  
hence the probability that a 5 or a 6 occurs at least once is  $1 - q^7 = \frac{2059}{2187}$ .

If we regard  $n$  and  $p$  as constant, then the above function  $P(k) = b(k; n, p)$  is a discrete probability distribution:

$k$	0	1	2	...	$n$
$P(k)$	$q^n$	$\binom{n}{1} q^{n-1} p$	$\binom{n}{2} q^{n-2} p^2$	...	$p^n$

It is called the *binomial distribution* since for  $k = 0, 1, 2, \dots, n$  it corresponds to the successive terms of the binomial expansion

$$(q + p)^n = q^n + \binom{n}{1} q^{n-1} p + \binom{n}{2} q^{n-2} p^2 + \dots + p^n$$

This distribution is also called the Bernoulli distribution, and independent trials with two outcomes are called Bernoulli trials.

Properties of this distribution follow:

**Theorem :**

Binomial distribution	
Mean	$\mu = np$
Variance	$\sigma^2 = npq$
Standard deviation	$\sigma = \sqrt{npq}$



**Example 3:** A fair die is tossed 180 times. The expected number of sixes is  $\mu = np = 180 \cdot \frac{1}{6} = 30$ . The standard deviation is  $\sigma = \sqrt{npq} = \sqrt{180 \cdot \frac{1}{6} \cdot \frac{5}{6}} = 5$ .

## 1. Example

Team A has probability  $\frac{2}{3}$  of winning whenever it plays. If A plays 4 games, find the probability that A wins (i) exactly 2 games, (ii) at least 1 game, (iii) more than half of the games.

Here  $n = 4$ ,  $p = \frac{2}{3}$  and  $q = 1 - p = \frac{1}{3}$ .

(i)  $P(2 \text{ wins}) = b(2; 4, \frac{2}{3}) = \binom{4}{2} (\frac{2}{3})^2 (\frac{1}{3})^2 = \frac{8}{27}$ .

(ii) Here  $q^4 = (\frac{1}{3})^4 = \frac{1}{81}$  is the probability that A loses all four games. Then  $1 - q^4 = \frac{80}{81}$  is the probability of winning at least one game.

(iii) A wins more than half the games if A wins 3 or 4 games. Hence the required probability is

$$P(3 \text{ wins}) + P(4 \text{ wins}) = \binom{4}{3} (\frac{2}{3})^3 (\frac{1}{3}) + \binom{4}{4} (\frac{2}{3})^4 = \frac{32}{81} + \frac{16}{81} = \frac{16}{27}$$

## 2.

A family has 6 children. Find the probability  $P$  that there are (i) 3 boys and 3 girls, (ii) fewer boys than girls. Assume that the probability of any particular child being a boy is  $\frac{1}{2}$ .

Here  $n = 6$  and  $p = q = \frac{1}{2}$ .

(i)  $P = P(3 \text{ boys}) = \binom{6}{3} (\frac{1}{2})^3 (\frac{1}{2})^3 = \frac{20}{64} = \frac{5}{16}$ .

(ii) There are fewer boys than girls if there are 0, 1 or 2 boys. Hence

$$P = P(0 \text{ boys}) + P(1 \text{ boy}) + P(2 \text{ boys}) = (\frac{1}{2})^6 + \binom{6}{1} (\frac{1}{2})(\frac{1}{2})^5 + \binom{6}{2} (\frac{1}{2})^2 (\frac{1}{2})^4 = \frac{11}{32}$$

## 3.



Determine the expected number of boys in a family with 8 children, assuming the sex distribution to be equally probable. What is the probability that the expected number of boys does occur?

The expected number of boys is  $E = np = 8 \cdot \frac{1}{2} = 4$ . The probability that the family has four boys is

$$b(4; 8, \frac{1}{2}) = \binom{8}{4} (\frac{1}{2})^4 (\frac{1}{2})^4 = \frac{8 \cdot 7 \cdot 6 \cdot 5}{1 \cdot 2 \cdot 3 \cdot 4} (\frac{1}{2})^8 = \frac{70}{256} = .27$$

4.

The probability is 0.02 that an item produced by a factory is defective. A shipment of 10,000 items is sent to its warehouse. Find the expected number  $E$  of defective items and the standard deviation  $\sigma$ .

$$E = np = (10,000)(0.02) = 200.$$

$$\sigma = \sqrt{npq} = \sqrt{(10,000)(0.02)(0.98)} = \sqrt{196} = 14.$$

## 6.2. Poisson Distribution

The Poisson distribution is defined as follows:

$$p(k; \lambda) = \frac{\lambda^k e^{-\lambda}}{k!}, \quad k = 0, 1, 2, \dots$$

where  $\lambda > 0$  is some constant. This countably infinite distribution appears in many natural phenomena, such as the number of telephone calls per minute at some switchboard, the number of misprints per page in a large text, and the number of  $\alpha$  particles emitted by a radioactive substance. Diagrams of the Poisson distribution for various values of  $\lambda$  follow.

Properties of the Poisson distribution follow:

**Theorem :**

Poisson distribution	
Mean	$\mu = \lambda$
Variance	$\sigma^2 = \lambda$
Standard deviation	$\sigma = \sqrt{\lambda}$



Although the Poisson distribution is of independent interest, it also provides us with a close approximation of the binomial distribution for small  $k$  provided that  $p$  is small and  $\lambda = np$ . This is indicated in the following table.

$k$	0	1	2	3	4	5
Binomial	.366	.370	.185	.0610	.0149	.0029
Poisson	.368	.368	.184	.0613	.0153	.00307

Comparison of Binomial and Poisson distributions  
with  $n = 100$ ,  $p = 1/100$  and  $\lambda = np = 1$

### Example

Suppose 2% of the items made by a factory are defective. Find the probability  $P$  that there are 3 defective items in a sample of 100 items.

The binomial distribution with  $n = 100$  and  $p = .02$  applies. However, since  $p$  is small, we use the Poisson approximation with  $\lambda = np = 2$ . Thus

$$P = p(3; 2) = \frac{2^3 e^{-2}}{3!} = 8(.135)/6 = .180$$

<Best Regards>