

# The Binomial Distribution

Definition. A binomial experiment consists of a sequence of  $n$  independent Bernoulli trials with  $P(S) = p$ .

The binomial random variable  $\underline{X}$  associated with this experiment is defined as

Notation:  $\underline{X} \sim \text{Bin}(n, p)$ .  $\underline{X}$  = the number of S's among  $n$  trials.

Question: We would like to know the probability  $p(k)$  of exactly  $k$  successes: (the pmf of  $\underline{X}$ )

$$p(k) = P(\underline{X} = k), \quad 0 \leq k \leq n, \quad \text{integer}$$

(Binomial distribution)

Notation: Because the pmf  $p(k)$ ,  $k = 0, 1, \dots, n$ , of  $\underline{X}$  depends on the two parameters  $n$  and  $p$ , we denote the pmf  $p(k)$  by  $b(k; n, p)$ .

Examples. 1°/ A coin is flipped  $n$  times. What is the chance of exactly  $k$  heads?

2°/ You have  $n$  chips. What is the chance that  $k$  are defective?

3°/ You buy  $n$  lottery scratch cards. What is the chance of  $k$  wins?

4°/ You type a page of  $n$  symbols. What is the chance of  $r$  errors?  
... ..

Theorem (Binomial distribution).  
of  $X \sim \text{Bin}(n, p)$

$$b(k; n, p) = \binom{n}{k} p^k (1-p)^{n-k}, \quad k=0, 1, 2, \dots, n$$

Proof: When we perform  $n$  Bernoulli trials there are exactly  $2^n$  possible outcomes, because each trial yields either S or F. How many of these outcomes comprise exactly  $k$  successes and  $n-k$  failures? The answer is

$$\binom{n}{k}$$

because this is the number of distinct ways of ordering  $k$  success and  $n-k$  failures. Now we observe that, by independence, any given outcome with  $k$  successes and  $n-k$  failures has probability  $p^k (1-p)^{n-k}$ .

Hence

$$b(k; n, p) = P(X=k) = \binom{n}{k} p^k (1-p)^{n-k}$$

$$= \left\{ \begin{array}{l} \text{number of outcomes} \\ \text{consisting of } k \text{ successes} \\ \text{(and } n-k \text{ failures)} \end{array} \right\} \times \left\{ \begin{array}{l} \text{probability of} \\ \text{any particular} \\ \text{such outcome} \end{array} \right\}.$$

Remark The Binomial distribution is indeed a probability distribution, because by the binomial theorem

$$\sum_{k=0}^n b(k; n, p) = \sum_{k=0}^n \binom{n}{k} p^k (1-p)^{n-k} = [p + (1-p)]^n = 1.$$

Properties of  $b(k; n, p)$

1°/ the recursion:

$$\begin{aligned} b(k+1; n, p) &= \binom{n}{k+1} p^{k+1} (1-p)^{n-k-1} \\ &= \frac{n-k}{k+1} \left( \frac{n!}{k!(n-k)!} \right) \left( \frac{p}{1-p} \right) p^k (1-p)^{n-k} \\ &= \frac{n-k}{k+1} \left( \frac{p}{1-p} \right) b(k; n, p) \end{aligned}$$

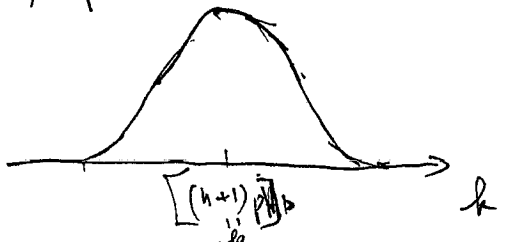
2°/ initialization

$$b(0; n, p) = (1-p)^n, \quad b(n; n, p) = p^n$$

Starting with  $b(0; n, p)$  or  $b(n; n, p)$ , we can use the recursive relation to carry out explicitly calculations.

3°/ shape of Binomial distribution.

histogram:  
 $b(k; n, p)$



Note that

$$\frac{b(k; n, p)}{b(k+1; n, p)} = \frac{k+1}{n-k} \left( \frac{1-p}{p} \right)$$

$$= \begin{cases} < 1, & \forall k < (n+1)p - 1 \\ > 1, & \forall k > (n+1)p + 1 \\ 1, & \text{if } k = \lfloor (n+1)p \rfloor \end{cases}$$

Binomial mean of  $X \sim \text{Bin}(n, p)$   $b(h; n, p)$

$$\mu = E(X) = \sum_{k=0}^n k \varphi(k) = \sum_{k=0}^n k \binom{n}{k} p^k (1-p)^{n-k}$$

$$= \sum_{k=1}^n k \frac{n!}{k!(n-k)!} p^k (1-p)^{n-k}$$

$$= n p \sum_{k=1}^n \frac{(n-1)!}{(k-1)!(n-k)!} p^{k-1} (1-p)^{n-k}$$

$$= n p \sum_{l=0}^{n-1} \frac{(n-1)!}{l!(n-1-l)!} p^l (1-p)^{n-1-l}$$

Binomial  
Theorem

$$\rightsquigarrow = n p [p + (1-p)]^{n-1}$$

$$= n p$$

$$\left[ \sum_{l=0}^{n-1} b(l; n-1, p) = 1 \right]$$

Moment generating function of  $X \sim \text{Bin}(n, p)$

$$M(t) = \sum_{k=0}^n e^{kt} \varphi(k) = \sum_{k=0}^n e^{kt} \binom{n}{k} p^k (1-p)^{n-k}$$

$$= \sum_{k=0}^n \binom{n}{k} (pe^t)^k (1-p)^{n-k}$$

$$= (pe^t + 1 - p)^n$$

Proof: The binomial theorem is:

$$(a + b)^n = \sum_{k=0}^n \binom{n}{k} a^k b^{n-k}$$

Note that

$$M(s) = \sum_{h=0}^n p(h) = (pe^{ts} + 1 - p)^n = 1.$$

The mean and variance of  $\bar{X}$  can be obtained by differentiating  $M(t)$ .

$$M'(t) = n(pe^{ts} + 1 - p)^{n-1} pe^{ts}$$

$$\Rightarrow \mu = M'(0) = np$$

$$M^{(2)}(t) = n(n-1)(pe^{ts} + 1 - p)^{n-2} (pe^{ts})(pe^{ts})$$

$$+ n(pe^{ts} + 1 - p)^{n-1} pe^{ts}$$

$$\Rightarrow M^{(2)}(0) = n(n-1)p^2 + np$$

$$\underset{\text{"}}{E(\bar{X}^2)}$$

Therefore

$$\sigma^2 = E(\bar{X}^2) - (E[\bar{X}])^2 = n(n-1)p^2 + np - n^2p^2$$

$$= np - np^2 = np(1-p).$$

Using Binomial Tables:

Appendix Table A.1 tabulates the cdf  $F(k) = P(\bar{X} \leq k)$

for  $n = 5, 10, 15, 20, 25$  in combination with selected values of  $\varphi$ . The cdf of  $\bar{X} \sim \text{Bin}(n, \varphi)$  is denoted by

$$h = 0, 1, \dots, n, \quad B(k; n, \varphi) = P(\bar{X} \leq k) = \sum_{l=0}^k b(l; n, \varphi)$$

Examples

Let  $\bar{X}$  = the nb of among 15 copies that failed the test with  $P(\text{failed}) = 0.2$ .

$$\sim \text{Bin}(n=15, \varphi=0.2)$$

1<sup>st</sup> Then the probability of at most 8 fail test is

$$P(\bar{X} \leq 8) = B(8; 15, 0.2) = 0.999$$

↑  
Table A.1

2<sup>nd</sup>. The probability of exactly 8 failed test is

$$\begin{aligned} P(\bar{X} = 8) &= P(\bar{X} \leq 8) - P(\bar{X} \leq 7) \\ &= B(8; 15, 0.2) - B(7; 15, 0.2) \\ &= 0.999 - 0.976 = 0.023. \end{aligned}$$