

## BINOMIAL DISTRIBUTION

Consider an experiment consisting of  $n$  independent "trials", each of which has 2 possible outcomes:

E.g. a series of  $n$  coin flips : H or T  
 ... dice rolls : 6 or not-6

Students taking a test : Pass/Fail

Molecules of a gas inside/outside  
 some volume  $v$  within a vessel.

The binomial distribution gives the probability of  $k$  "successes" in  $n$  trials.

EXAMPLE 1 : # of heads in 3 coin tosses

Outcomes	#heads, $k$	$P_k$
TTT	0	$(\frac{1}{2})^3$
TTH, THT, HTT	1	$3 \times (\frac{1}{2})^2 \times (\frac{1}{2})$
THH, HTH, HHT	2	$3 \times (\frac{1}{2}) \times (\frac{1}{2})^2$
HHH	3	$(\frac{1}{2})^3$

## [BINOMIAL DIST, CONTINUED]

EXAMPLE 2 : # of sixes in 2 dice rolls

Outcomes	# of sixes, $k$	$P_k$
XX	0	$(\frac{5}{6})^2$
6X, X6	1	$2 \times \frac{1}{6} \times \frac{5}{6}$
66	2	$(\frac{1}{6})^2$

GENERAL CASE : Each trial has probability

$p$  of "success" and  $q$  [ $\equiv 1-p$ ] of "failure". One possibility is that we have  $k$  successes followed by  $(n-k)$  failures, with prob.  $\underbrace{p \times p \times \dots \times p}_{k \text{ times}} \times \underbrace{q \times \dots \times q}_{n-k \text{ times}}$

But the successes and failures could occur in any one of  $\frac{n!}{k!(n-k)!} = \binom{n}{k}$  ways,

so the total probability of  $k$  successes will be

$$P_k = \binom{n}{k} p^k q^{n-k} \quad \text{BINOMIAL DISTRIBUTION}$$

# [BINOMIAL DISTRIBUTION, CONT.]

In EXAMPLE 1,  $n = 3$  and  $p = q = \frac{1}{2}$ .

Then, for example,  $P_2 = \left(\frac{3}{2}\right) \times \left(\frac{1}{2}\right)^2 \times \left(\frac{1}{2}\right) = \frac{3}{8}$ .

In EXAMPLE 2,  $n = 2$  and  $p = \frac{1}{6}$ ,  $q = \frac{5}{6}$

$$\Rightarrow P_1 = \left(\frac{2}{1}\right) \times \frac{1}{6} \times \frac{5}{6} = \frac{5}{12}.$$

CHECK: As always for a probability distribution, the probabilities must add up to 1.

$$\sum_{k=0}^n P_k = \sum_{k=0}^n \binom{n}{k} p^k q^{n-k}$$

$$= (p+q)^n \text{ [BINOMIAL THEOREM]}$$

$$= 1 \text{ [since } p+q = 1\text{]}$$

We often need to know the expected [mean] number of successes  $\langle k \rangle$  and the variance  $\sigma_k^2 = \text{var}(k) = \langle k^2 \rangle - \langle k \rangle^2$ .

## MEAN OF BINOMIAL DIST. :

Expect this to be  $np$ , as the relative frequency of success in repeated single trials is  $p$ .

$$\begin{aligned} \text{CHECK: } \langle k \rangle &= \sum_{k=0}^n k P_k \\ &= \sum_{k=0}^n k \binom{n}{k} p^k q^{n-k} \\ &= q^n \sum_{k=0}^n \binom{n}{k} k x^k \end{aligned}$$

$$\text{where } x = \frac{p}{q}$$

$$\text{But } kx^k = x \frac{d}{dx} (x^k),$$

$$\Rightarrow \langle k \rangle = q^n x \frac{d}{dx} \sum_{k=0}^n \binom{n}{k} x^k 1^{n-k}$$

$$= q^n x \frac{d}{dx} (x+1)^n \text{ BINOMIAL THEOREM}$$

$$= q^n x \cdot n (x+1)^{n-1}$$

$$= q^n \cdot \frac{p}{q} \cdot n \left(\frac{p}{q} + 1\right)^{n-1}$$

$$= p \cdot n (p+q)^{n-1}$$

$$= np \text{ (since } p+q = 1\text{)}$$

## VARIANCE OF BINOMIAL DIST.:

We follow the same approach that we used with the Poisson distribution, i.e., look at

$$\begin{aligned}\langle k(k-1) \rangle &= \langle k^2 \rangle - \langle k \rangle \\ &= [\sigma_k^2 + \langle k \rangle^2] - \langle k \rangle.\end{aligned}$$

Details:

$$\begin{aligned}\langle k(k-1) \rangle &= \sum_{k=0}^n k(k-1) \binom{n}{k} p^k q^{n-k} \\ &= q^n \sum_{k=0}^n \binom{n}{k} \underbrace{k(k-1) x^k}_{x^2 \frac{d^2}{dx^2} x^k} \\ &= q^n x^2 \frac{d^2}{dx^2} \sum_{k=0}^n \binom{n}{k} x^k \\ &= q^n x^2 \cdot n(n-1) (x+1)^{n-2} \\ &= p^2 \cdot n(n-1) (p+q)^{n-2} \\ &= n(n-1) p^2\end{aligned}$$

Hence

$$\begin{aligned}\sigma_k^2 &= n(n-1) p^2 - \langle k \rangle^2 + \langle k \rangle \\ &= (n^2 p^2 - n p^2) - n^2 p^2 + n p \\ &= n(p - p^2) = n p q.\end{aligned}$$

## MEAN & VARIANCE, CONT'D

<p>SUMMARY: <math>\langle k \rangle = np</math>  <math>\sigma_k^2 = npq</math></p>
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Note that mean is proportional to  $n$  and the "width" of the distribution,  $\sigma_k$ , varies as  $\sqrt{n}$  so that "fractional width"  $\sigma_k / \langle k \rangle \propto \frac{1}{\sqrt{n}}$  decreases with increasing  $n$ : the binomial distribution becomes [relatively] sharply peaked about its mean.

For large  $n$ ,  $P_k$  can be approximated by a GAUSSIAN DISTRIBUTION of mean  $np$  and s.d.  $(npq)^{1/2}$ .

[Proof is similar to large  $\lambda$  limit of Poisson, with Stirling's approx. used for each of the factorials in the binomial coefficients  $\frac{n!}{k!(n-k)!}$ .]

**EXAMPLE 3:** Each of the 3 lifts in the Schuster lab may fail with probability 0.1 on a given day. Calculate:

- (a) the probability that only one lift is working all day;
- (b) the probability that at least one lift fails;
- (c) the mean number of failed lifts.

**SOLUTION:**  $n=3$ ,  $p=0.1$ ,  $q=0.9$

$$\begin{aligned} \text{(a)} \quad P_2 &= \binom{3}{2} p^2 q \\ &= 3 \times 0.1^2 \times 0.9 = 0.027 \end{aligned}$$

$$\begin{aligned} \text{(b)} \quad P_1 + P_2 + P_3 &= 1 - P_0 \\ &= 1 - (0.9)^3 = 0.271 \end{aligned}$$

$$\text{(c)} \quad np = 3 \times 0.1 = 0.3$$