

## Some Discrete Probability Distributions

### The Bernoulli Process

An experiment often consists of repeated trials, each with two possible outcomes that may be labeled success or failure.

Eg: Tossing a coin ten times and finding the probability of number of heads.

Head — Success

Tail — Failure.

The process is referred to as a Bernoulli process. Each trial is called a Bernoulli trial.

The Bernoulli ~~trial~~ process must possess the following properties:

1. The experiment consists of repeated trials.
2. Each trial results in an outcome that may be classified as a success or a failure.
3. The probability of success, denoted by  $p$ , remains constant from trial to trial.
4. The repeated trials are independent.

Note: In drawing cards from a deck, the probabilities for repeated trials change if the card is not replaced.

The probability of selecting a heart on first draw is  $\frac{13}{52}$  and on second draw it is conditional probability,  $\frac{13}{51}$  or  $\frac{12}{51}$

depending on whether a heart appeared on first draw.

$\therefore$  This would not be considered as a Bernoulli trial.

## Binomial Distribution

The number  $X$  of successes in  $n$  Bernoulli trials is called a binomial random variable. The probability distribution of this discrete random variable is called the binomial distribution and its value is denoted by  $b(x; n, p)$ .

Def. A Bernoulli trial can result in a success with probability  $p$  and a failure with probability  $q = 1 - p$ . Then the probability distribution of the binomial random variable  $X$ , the number of successes in  $n$  independent trials, is

$$b(x; n, p) = {}^nC_x p^x q^{n-x}, \quad x = 0, 1, 2, \dots, n.$$

$${}^nC_x = \binom{n}{x}$$

Ex. Three items are selected at random from a manufacturing process, inspected and classified as defective or nondefective. Find the probability distribution for number of defectives assuming that 25% items are defective.

Sol. Let  $X$  be a random variable representing number of defectives.  $P(S) = p = \frac{1}{4}$ ,  $q = \frac{3}{4}$

$$S = \{NNN, NDN, NND, DNN, NDD, DND, DDN, DDD\}$$

$X$	0	1	2	3
$f(x)$	$\frac{27}{64}$	$\frac{27}{64}$	$\frac{9}{64}$	$\frac{1}{64}$

$$f(0) = P(NNN) = \frac{3}{4} \cdot \frac{3}{4} \cdot \frac{3}{4} = \frac{27}{64}$$

$$P(1) = 3 \cdot \frac{1}{4} \cdot \frac{3}{4} \cdot \frac{3}{4} = \frac{27}{64}$$

$$f(2) = 3 \cdot \frac{1}{4} \cdot \frac{1}{4} \cdot \frac{3}{4} = \frac{9}{64}$$

$$f(3) = \frac{1}{4} \cdot \frac{1}{4} \cdot \frac{1}{4} = \frac{1}{64}$$

$$b(x; 3, \frac{1}{4}) = {}^3C_x \left(\frac{1}{4}\right)^x \left(\frac{3}{4}\right)^{3-x}, \quad n = 0, 1, 2, 3.$$



Ex.: ~~Five~~ <sup>Ten</sup> coins are thrown simultaneously. Find the probability of getting at least seven heads.

Sol.:  $p = \text{Probability of getting a head} = \frac{1}{2}$   
 $q = \text{Probability of not getting a head} = \frac{1}{2}$

The probability of getting  $x$  heads in a random throw of 10 coins is

$$b(x; 10, \frac{1}{2}) = {}^{10}C_x \left(\frac{1}{2}\right)^x \left(\frac{1}{2}\right)^{10-x}, x = 0, 1, 2, \dots, 10$$

Probability of getting at least 7 heads

$$\begin{aligned} &= P(X \geq 7) = P(X=7) + P(X=8) + P(X=9) + P(X=10) \\ &= {}^{10}C_7 \left(\frac{1}{2}\right)^7 \left(\frac{1}{2}\right)^3 + {}^{10}C_8 \left(\frac{1}{2}\right)^8 \left(\frac{1}{2}\right)^2 + {}^{10}C_9 \left(\frac{1}{2}\right)^9 \left(\frac{1}{2}\right)^1 \\ &\quad + {}^{10}C_{10} \left(\frac{1}{2}\right)^{10} \left(\frac{1}{2}\right)^0 \end{aligned}$$

$$= \left(\frac{1}{2}\right)^{10} [{}^{10}C_7 + {}^{10}C_8 + {}^{10}C_9 + {}^{10}C_{10}]$$

$$= \frac{1}{1024} \left[ \frac{10!}{7! 3!} + \frac{10!}{8! 2!} + \frac{10!}{9! 1!} + \frac{10!}{0! 10!} \right]$$

$$= \frac{1}{1024} \left[ \frac{10 \cdot 9 \cdot 8}{3 \cdot 2} + \frac{10 \cdot 9}{2} + 10 + 1 \right]$$

$$= \frac{1}{1024} [120 + 45 + 11]$$

$$= \frac{176}{1024}$$

Ex The probability that a certain kind of component will survive a shock test is  $\frac{3}{4}$ . Find the probability that exactly 2 of the next 4 components tested survive.

Sol :  $p = \frac{3}{4}, q = \frac{1}{4}$

$$b\left(2; 4, \frac{3}{4}\right) = {}^4C_2 \left(\frac{3}{4}\right)^2 \left(\frac{1}{4}\right)^2 = 6 \cdot \frac{9}{16} \cdot \frac{1}{16} = \frac{27}{128}.$$

Ex The probability that a patient recovers from a rare blood disease is 0.4. If 15 people are known to have contracted this disease, what is the probability that

- (a) at least 10 survive
- (b) from 3 to 8 survive
- (c) exactly 5 survive?

$t) = E[$  Moment Generating Function (m.g.f.).

The m.g.f. of a RV  $X$  (about origin) having the probability function  $f(x)$  is given by

$$M_X(t) = E(e^{tx}) = \begin{cases} \int e^{tx} f(x) dx, & \text{for continuous probability distribution} \\ \sum e^{tx} f(x), & \text{for discrete probability distribution} \end{cases}$$

$$\begin{aligned} M_X(t) &= E(e^{tx}) = E\left(1 + tx + \frac{(tx)^2}{2!} + \dots + \frac{(tx)^r}{r!} + \dots\right) \\ &= 1 + tE(X) + \frac{t^2}{2!} E(X^2) + \dots + \frac{t^r}{r!} E(X^r) \\ &\quad + \dots \end{aligned}$$

$$\begin{aligned} &= 1 + t\mu_1' + \frac{t^2}{2!} \mu_2' + \dots + \frac{t^r}{r!} \mu_r' + \dots \\ &= \sum_{r=0}^{\infty} \frac{t^r}{r!} \mu_r' \end{aligned}$$

where  $\mu_r' = E(X^r) = \begin{cases} \int_{-\infty}^{\infty} x^r f(x) dx, & \text{for continuous distribution} \\ \sum_x x^r f(x), & \text{for discrete distribution,} \end{cases}$

is the  $r$ th moment of  $X$  about origin.

Thus,  $\mu_r'$  (about origin) = Coefficient of  $\frac{t^r}{r!}$  in  $M_X(t)$ .

Since,  $M_X(t)$  generates moments, it is known as m.g.f.



$$\text{or } \mu'_2 = \left. \frac{d^2}{dt^2} \{M_X(t)\} \right|_{t=0}$$

In general, the mgf of  $X$  about the point  $X=a$  is defined as:

$$\begin{aligned} M_X(t) \text{ (about } a) &= E[e^{t(X-a)}] \\ &= E\left[1 + t(X-a) + \frac{t^2(X-a)^2}{2!} + \dots \right. \\ &\quad \left. + \frac{t^r}{r!} (X-a)^r + \dots\right] \\ &= 1 + t\mu'_1 + \frac{t^2}{2!} \mu'_2 + \dots + \frac{t^r}{r!} \mu'_r + \dots \end{aligned}$$

where  $\mu'_r = E[(X-a)^r]$ , is the  $r$ th moment about the point  $X=a$ .

Find the Moment Generating Function of Binomial Distribution and use it to find the  $\mu$  and  $\sigma^2$

Let  $X$  be a binomial random variable.

$$\therefore b(x; n, p) = {}^n C_x p^x q^{n-x}, \quad x=0, 1, 2, \dots, n.$$

$$M_X(t) = E(e^{tx}) = \sum_{x=0}^n e^{tx} {}^n C_x p^x q^{n-x}$$

$$= \sum_{x=0}^n {}^n C_x (pe^t)^x q^{n-x} = (q + pe^t)^n$$

$$\left[ \text{Binomial Expansion: } (a+x)^n = {}^n C_0 a^n x^0 + {}^n C_1 a^{n-1} x^1 + \dots + {}^n C_n a^0 x^n \right]$$

## Mean and Variance of Binomial Distribution

$\mu = E(X) = \mu_1'$ , the first moment about origin.

$$\sigma^2 = E(X^2) - [E(X)]^2 = \mu_2' - (\mu_1')^2,$$

Where  $\mu_2'$  is the second moment about origin.

$$\text{Now, } E(X^2) = \left. \frac{d^2}{dt^2} [M_X(t)] \right|_{t=0}$$

$$\therefore E(X) = \left. \frac{d}{dt} [M_X(t)] \right|_{t=0} = \left. n(q + pet)^{n-1} pet \right|_{t=0}$$

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

$$= np \quad [\because q+p=1]$$

$$\therefore \boxed{\mu = E(X) = np}$$

$$\Rightarrow E(X^2) = \left. \frac{d^2}{dt^2} (M_X(t)) \right|_{t=0}$$

$$= \left. \frac{d}{dt} (npet(q+pet)^{n-1}) \right|_{t=0}$$

$$= \left. np \left\{ e^{t(n-1)} (q+pet)^{n-2} pet + e^t (q+pet)^{n-1} \right\} \right|_{t=0}$$

$$= np[(n-1)(q+p)p + (q+p)^{n-1}]$$

$$= np[np - p + 1] = n^2p^2 - np^2 + np$$

$$\begin{aligned}\sigma^2 &= n^2 p^2 - np^2 + np - n^2 p^2 - \cancel{np(np)} \\ &= np(1-p) = npq\end{aligned}$$

$$\boxed{\sigma^2 = npq}$$



# Negative Binomial and Geometric Distributions

## Negative Binomial Experiments

Consider an experiment where the properties are the same as those listed for a binomial experiment, with the exception that the trials will be repeated until a fixed number of successes occur.

Therefore, instead of the probability of  $x$  successes in  $n$  trials, where  $n$  is fixed, we are now interested in the probability that the  $k$ th success occurs on the  $x$ th trial. Experiments of this kind are called negative binomial experiments.

## Negative Binomial Random Variable

The number  $X$  of trials required to produce  $k$  successes in a negative binomial experiment is called a negative binomial random variable and its probability distribution is called the negative binomial distribution.

## Negative Binomial distribution

If repeated independent trials can result in a success with probability  $p$  and a failure with probability  $q = 1 - p$ , then the probability distribution of the random variable  $X$ , the number of the trial on which the  $k$ th success occurs, is

$$b^*(x; k, p) = \binom{x-1}{k-1} p^k q^{x-k}, \quad x = k, k+1, \dots$$

1. Q: Find the probability that a person flipping a coin gets
- (a) the third head on the seventh flip
  - (b) the first head on the fourth flip.

$$\text{Sol: (a) } b^*(7, 3, \frac{1}{2}) = {}^6C_3 \left(\frac{1}{2}\right)^3 \left(\frac{1}{2}\right)^4$$

$$= \frac{15}{128} = 0.1172$$

$$\cancel{b^*(4, 1, \frac{1}{2})} \quad (b) \quad b^*(4, 1, \frac{1}{2}) = {}^3C_0 \left(\frac{1}{2}\right)^1 \left(\frac{1}{2}\right)^3 = \frac{1}{16} = 0.0625$$

Ex In an NBA (National Basketball Association) championship series, the team that wins four games out of seven is the winner. Suppose that the teams A and B face each other in the championship games and that team A has probability 0.55 of winning a game over team B.

- (a) What is the probability that team A will win the series in 6 games?
- (b) What is the probability that team A will <sup>win</sup> the series? ~~in 6 games?~~
- (c) If teams A and B were facing each other in a regional playoff series, which is decided by winning three out of five games, what is the probability that team A would win the series?

$$\text{Sol: (a) } b^*(6; 4) 0.55 = {}^5C_3 (0.55)^4 (0.45)^2$$

$$= 0.1853$$

$$(b) \quad P(\text{A will win the series})$$

$$= P(X \geq 4)$$

$$= b^*(4, 4, 0.55) + b^*(5, 4, 0.55) + b^*(6, 4, 0.55) + b^*(7, 4, 0.55)$$

$$= {}^3C_3 (0.55)^4 (0.45)^0 + {}^4C_3 (0.55)^4 (0.45)^1 + {}^5C_3 (0.55)^4 (0.45)^2 +$$

$${}^6C_3 (0.55)^4 (0.45)^3$$

$$= 0.0915 + 0.1647 + 0.1853 + 0.1668$$

$$= 0.6083$$



(C)  $P(\text{team A wins the playoff})$

$$= P(X \geq 3)$$

$$= b^*(3; 3, 0.55) + b^*(4; 3, 0.55) + b^*(5; 3, 0.55)$$

$$= {}^3C_2 (0.55)^3 (0.45)^0 + {}^3C_2 (0.55)^2 (0.45)^1 + {}^4C_2 (0.55)^3 (0.45)^2$$

$$= 0.1664 + 0.2246 + 0.2021$$

$$= 0.5931$$

If we consider the special case of the negative binomial distribution where  $k=1$ , we have a probability distribution for the number of trials required for a single success.

For eg: Tossing a coin until head occurs.

We might be interested in the probability that the first head occurs on the fourth toss.

The negative binomial reduces to

$$b^*(x; 1, p) = pq^{x-1}, \quad x=1, 2, 3, \dots$$

### Geometric Distribution

If repeated independent trials can result in a success with probability  $p$  and a failure with probability  $q=1-p$ , then the probability distribution of the random variable  $X$ , the number of the trial on which the first success occurs, is

$$g(x; p) = pq^{x-1}, \quad x=1, 2, 3, \dots$$

$$1 \text{ a (b)} \quad g(4; \frac{1}{2}) = (\frac{1}{2}) (\frac{1}{2})^3 = \frac{1}{2^4} = \frac{1}{16} = 0.0625.$$

Ex For a certain manufacturing process, it is known that, on the average, 1 in every 100 items is defective. What is the probability that the fifth item inspected is

the first defective item found?

Sol:-  $g(5; 0.01) = (0.01)(0.99)^4$   
 $= 0.0096$

$$p = \frac{1}{100} = 0.01$$
$$q = 0.99$$

Ex:- At a busy time, a telephone exchange is very near capacity, so callers have difficulty placing their calls. It may be of interest to know the number of attempts necessary in order to make a connection. Suppose that we let  $p = 0.05$  be the probability of a connection during a busy time. We are interested in knowing the probability that 5 attempts are necessary for a successful call.

Sol:-  $g(5; 0.05) = (0.05)(0.95)^4 = 0.0407.$

Mean and Variance of a random variable following the Geometric Distribution:-

$$\mu = \frac{1}{p}, \quad \sigma^2 = \frac{q}{p^2}.$$

Q:- In the above example, find the expected number of calls necessary to make a connection?

Sol:-  $\mu = \frac{1}{0.05} = 20.$



## Poisson Process and Poisson Distribution

Some experiments results in counting the number of particular events occur in given time interval or in a specified region, known as Poisson Experiments.

The time interval may be of any length, such as a minute, a day, a week, a month or even a year.

Eg-1. 1. ~~Ex~~ Number of telephone calls received per hour by an office.

2. How many vehicles pass through a traffic signal in a day.

3. How many people arrive at a railway station from 9 am to 11 am.

4. How many people enter in the door of a shopping mall in January.

The specified region can be a line segment, <sup>on</sup> area, a volume or perhaps a piece of material.

Eg-1. Number of field mice per acre.

2. Number of typing errors per page.

## Poisson Process

Poisson Process represents observations / occurrences / happenings over time / area.

## Properties of Poisson Process

1. The number of outcomes / occurrences during disjoint time intervals are **independent** (accidents)

Eg.: No. of earthquakes recorded in 2021-22 is independent of the no. of earthquakes recorded in 2001-02. (accidents)

2. The probability of a single occurrence during a small time interval is proportional to the length of the interval.  
 $P_1(h) = P(X(h)=1) = \lambda h$  (Rate of occurrence of an event)

3. The probability of more than one occurrence during a small time interval is negligible.

Eg.: If there is a train accident at 9.00 am at a particular place, then it is highly ~~likely~~ unlikely that there will be a train accident at 9.03 am.

### Poisson Random Variable and Poisson Distribution

The number  $X$  of outcomes occurring during a Poisson experiment is called a Poisson Random Variable and its probability distribution is called the Poisson distribution.

### Poisson Distribution

Def.: The probability distribution of the Poisson random variable  $X$ , representing the number of outcomes occurring in a given time interval or specified region denoted by  $t$ , is

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

where  $\lambda$  is the average number of outcomes per unit time, distance, area or volume and  $e = 2.71828$ .

Ex: During a laboratory experiment, the average number of radioactive particles passing through a counter in 1 millisecond is 4. What is the probability that 6 particles enter the counter in a given millisecond?



Sol  $\therefore \lambda t = 4, x = 6$

$$p(6; 4) = \frac{e^{-4}(4)^6}{6!} = \frac{(0.0183)(4096)}{720} = \frac{74.9568}{720} = 0.1041$$

Mean and Variance of Poisson Distribution  $p(x; \lambda t)$

$$\boxed{\mu = \lambda t, \sigma^2 = \lambda t}$$

$$\begin{aligned} \mu = E(X) &= \sum_{x=0}^{\infty} x \frac{e^{-\lambda t} (\lambda t)^x}{x!} = \sum_{x=1}^{\infty} \frac{x e^{-\lambda t} (\lambda t)^x}{x!} \\ &= e^{-\lambda t} (\lambda t) \sum_{x=1}^{\infty} \frac{(\lambda t)^{x-1}}{(x-1)!} \\ &= e^{-\lambda t} (\lambda t) \left[ 1 + \frac{\lambda t}{1} + \frac{(\lambda t)^2}{2!} + \frac{(\lambda t)^3}{3!} + \dots \right] \\ &= \lambda t e^{-\lambda t} e^{\lambda t} \end{aligned}$$

$$\boxed{\mu = \lambda t}$$

$$\begin{aligned} \sigma^2 &= E(X^2) - \mu^2 \\ &= \sum_{x=0}^{\infty} x^2 \frac{e^{-\lambda t} (\lambda t)^x}{x!} - \lambda^2 t^2 \\ &= \sum_{x=1}^{\infty} (x^2 - x + x) \frac{e^{-\lambda t} (\lambda t)^x}{x!} - \lambda^2 t^2 \end{aligned}$$

$$= \sum_{x=2}^{\infty} \frac{x(x-1) e^{-\lambda t} (\lambda t)^x}{x(x-1)(x-2)!} + \lambda t - \lambda^2 t^2$$

$$= e^{-\lambda t} (\lambda t)^2 \sum_{x=2}^{\infty} \frac{(\lambda t)^{x-2}}{(x-2)!} + \lambda t - \lambda^2 t^2$$

$$= e^{-\lambda t} (\lambda t)^2 \left[ 1 + \frac{\lambda t}{1!} + \frac{(\lambda t)^2}{2!} + \frac{(\lambda t)^3}{3!} + \dots \right] + \lambda t - \lambda^2 t^2$$

$$= e^{-\lambda t} (\lambda t)^2 e^{\lambda t} + \lambda t - \lambda^2 t^2$$

$$= \lambda^2 t^2 + \lambda t - \lambda^2 t^2$$

$$= \lambda t$$

$$\boxed{\sigma^2 = \lambda t}$$

## Approximation of Binomial Distribution by a Poisson Distribution

Poisson distribution is a limiting case of the binomial distribution under the following conditions:

- (i)  $n$ , the number of trials is indefinitely large, i.e.,  $n \rightarrow \infty$
- (ii)  $p$ , the constant probability of success for each trial is indefinitely small, i.e.,  $p \rightarrow 0$ .
- (iii)  $np = \lambda$ ,  $np = \mu$ , is finite.

Thm Let  $X$  be a binomial random variable with probability distribution  $b(x; n, p)$ . When  $n \rightarrow \infty$ ,  $p \rightarrow 0$  and  $np \xrightarrow{n \rightarrow \infty} \mu$  remains constant,

$$b(x; n, p) \xrightarrow{n \rightarrow \infty} p(x; \mu).$$

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



Ex In a certain industrial facility, accidents occur infrequently. It is known that the probability of an accident on any given day is 0.005 and accidents are independent of each other.

- Sol: (a) What is the probability that in any given period of 400 days there will be an accident on one day?
- (b) What is the probability that there are at most three days with accidents?

Sol: Let  $X$  be a binomial random variable with  $n=400$  and  $p=0.005$ .

$$\text{Thus, } np = 400 \times 0.005 = 2$$

Using Poisson Process,

$$(a) P(X=1) = \frac{e^{-2} 2^1}{1!} = (0.1353)(2) = 0.2706$$

$$\begin{aligned} (b) P(X \leq 3) &= P(X=1) + P(X=2) + P(X=3) + P(X=0) \\ &= 0.2706 + \frac{e^{-2} 2^2}{2!} + \frac{e^{-2} 2^3}{3!} + \frac{e^{-2} 2^0}{0!} \\ &= 0.2706 + \frac{(0.1353)(4)}{2} + \frac{(0.1353)(8)}{6} + 0.1353 \\ &= 0.2706 + 0.2706 + 0.1804 + 0.1353 \\ &= 0.7216 + 0.1353 \\ &= 0.8569 \end{aligned}$$

Ex In a manufacturing process where glass products are made, defects or bubbles occur, occasionally rendering the piece undesirable for marketing. It is known that, on average, 1 in every 1000 of these items produced has one or more bubbles. What is the probability that a

random sample of 8000 will yield fewer than 7 items possessing bubbles?

Sol: It is a binomial experiment with  $n=8000$  and  $p=0.001$ . Since  $p$  is very close to 0 and  $n$  is quite large, we will use Poisson distribution.

$$\mu = 8000 \times 0.001 = 8.$$

Let  $X$  represent the number of bubbles.

$$P(X < 7) = P(X=0) + P(X=1) + P(X=2) + P(X=3) + P(X=4) + P(X=5) + P(X=6)$$

$$= e^{-8} \left[ \frac{8^0}{0!} + \frac{8^1}{1!} + \frac{8^2}{2!} + \frac{8^3}{3!} + \frac{8^4}{4!} + \frac{8^5}{5!} + \frac{8^6}{6!} \right]$$

$$= e^{-8} [1 + 8 + 32 + 85.3333 + 170.6667 + 273.0667 + 364.0889]$$

$$= \cancel{0.00033} [934.1556] \times 0.0003355$$

$$= 0.3134$$

Ex A manufacturer, who produces medicine bottles, finds that 0.1% of the bottles are defective. The bottles are packed in boxes containing 500 bottles. A drug manufacturer buys 100 boxes from the producer of bottles. Using Poisson distribution, find how many boxes will contain:

(i) no defective (ii) at least two defectives.

Sol:  $n=500$ ,  $p=0.001$ ,  $np=0.5$

Let  $X$  be a random variable denote the number of defective bottles in a box of 500. The prob of  $x$  defective bottle in a box is

$$\therefore P(X=x) = \frac{e^{-0.5} (0.5)^x}{x!}, \quad x=0, 1, 2, \dots$$



∴ The number of boxes containing  $x$  defective bottles in a consignment of 100 boxes is

$$100 \times P(X=x) = 100 \times \frac{e^{-0.5} \times (0.5)^x}{x!}, \quad x=0, 1, 2, \dots$$

(i) Number of boxes containing no defective bottles is

$$\begin{aligned} 100 \times P(X=0) &= 100 \times \frac{e^{-0.5} \times (0.5)^0}{0!} \\ &= 100 \times 0.6065 \\ &= 60.65 \\ &\approx 61. \end{aligned}$$

(ii) Number of boxes containing at least two defective bottles

$$\begin{aligned} \text{is } 100 \times P(X \geq 2) &= 100 [1 - P(X < 2)] \\ &= 100 [1 - P(X=0) - P(X=1)] \\ &= 100 \left[ 1 - 0.6065 \times 1 - \frac{0.6065 \cdot (0.5)}{1!} \right] \\ &= 100 [1 - 0.6065 - 0.3033] \\ &= 100 [0.0902] \\ &= 9.02 \\ &\approx 9. \end{aligned}$$

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$$\begin{aligned} M_X(t) &= E(e^{tx}) = E\left(1 + tx + \frac{(tx)^2}{2!} + \dots + \frac{(tx)^r}{r!} + \dots\right) \\ &= 1 + tE(X) + \frac{t^2}{2!} E(X^2) + \dots + \frac{t^r}{r!} E(X^r) \\ &\quad + \dots \end{aligned}$$

$$\begin{aligned} &= 1 + t\mu_1' + \frac{t^2}{2!} \mu_2' + \dots + \frac{t^r}{r!} \mu_r' + \dots \\ &= \sum_{r=0}^{\infty} \frac{t^r}{r!} \mu_r' \end{aligned}$$

where  $\mu_r' = E(X^r) = \begin{cases} \int_{-\infty}^{\infty} x^r f(x) dx, & \text{for continuous distribution} \\ \sum_x x^r f(x), & \text{for discrete distribution,} \end{cases}$

is the  $r$ th moment of  $X$  about origin.

Thus,  $\mu_r'$  (about origin) = Coefficient of  $\frac{t^r}{r!}$  in  $M_X(t)$ .

Since,  $M_X(t)$  generates moments, it is known as m.g.f.



$$\text{or } \mu'_2 = \left. \frac{d^2}{dt^2} \{M_X(t)\} \right|_{t=0}$$

In general, the mgf of  $X$  about the point  $X=a$  is defined as:

$$\begin{aligned} M_X(t) \text{ (about } a) &= E[e^{t(X-a)}] \\ &= E\left[1 + t(X-a) + \frac{t^2(X-a)^2}{2!} + \dots \right. \\ &\quad \left. + \frac{t^r}{r!}(X-a)^r + \dots\right] \\ &= 1 + t\mu'_1 + \frac{t^2}{2!}\mu'_2 + \dots + \frac{t^r}{r!}\mu'_r + \dots \end{aligned}$$

where  $\mu'_r = E[(X-a)^r]$ , is the  $r$ th moment about the point  $X=a$ .

Find the Moment Generating Function of Binomial Distribution and use it to find the  $\mu$  and  $\sigma^2$

Let  $X$  be a binomial random variable.

$$\therefore b(x; n, p) = {}^n C_x p^x q^{n-x}, \quad x=0, 1, 2, \dots, n.$$

$$M_X(t) = E(e^{tx}) = \sum_{x=0}^n e^{tx} {}^n C_x p^x q^{n-x}$$

$$= \sum_{x=0}^n {}^n C_x (pe^t)^x q^{n-x} = (q + pe^t)^n$$

$$\left[ \text{Binomial Expansion: } (a+x)^n = {}^n C_0 a^n x^0 + {}^n C_1 a^{n-1} x^1 + \dots + {}^n C_n a^0 x^n \right]$$

## Mean and Variance of Binomial Distribution

$\mu = E(X) = \mu_1'$ , the first moment about origin.

$$\sigma^2 = E(X^2) - [E(X)]^2 = \mu_2' - (\mu_1')^2,$$

Where  $\mu_2'$  is the second moment about origin.

$$\text{Now, } E(X^2) = \left. \frac{d^2}{dt^2} [M_X(t)] \right|_{t=0}$$

$$\therefore E(X) = \left. \frac{d}{dt} [M_X(t)] \right|_{t=0} = \left. n(q + pet)^{n-1} pet \right|_{t=0}$$

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

$$= np \quad [\because q+p=1]$$

$$\therefore \boxed{\mu = E(X) = np}$$

$$\Rightarrow E(X^2) = \left. \frac{d^2}{dt^2} (M_X(t)) \right|_{t=0}$$

$$= \left. \frac{d}{dt} (npet(q+pet)^{n-1}) \right|_{t=0}$$

$$= \left. np \left\{ e^{t(n-1)} (q+pet)^{n-2} pet + e^t (q+pet)^{n-1} \right\} \right|_{t=0}$$

$$= np[(n-1)(q+p)p + (q+p)^{n-1}]$$

$$= np[np - p + 1] = n^2p^2 - np^2 + np$$



$$\begin{aligned}\sigma^2 &= n^2 p^2 - np^2 + np - n^2 p^2 - \cancel{np(np)} \\ &= np(1-p) = npq\end{aligned}$$

$$\boxed{\sigma^2 = npq}$$

## M. g.f. of Negative Binomial Distribution

$$\begin{aligned}M_X(t) &= E(e^{tx}) = \sum_{x=k}^{\infty} e^{tx} \binom{x-1}{k-1} q^{x-k} p^k \\&= p^k e^{tk} \sum_{x=k}^{\infty} e^{t(x-k)} \binom{x-1}{k-1} q^{x-k} \\&= p^k e^{tk} \sum_{x=k}^{\infty} \binom{x-1}{k-1} (qe^t)^{x-k} \\&= p^k e^{tk} \sum_{r=0}^{\infty} \binom{k+r-1}{k-1} (qe^t)^r\end{aligned}$$

$$\begin{aligned}\text{let } r &= x-k \\ \Rightarrow x &= k+r\end{aligned}$$

$$\begin{aligned}&= (pe^t)^k \left[ 1 + k(qe^t) + \frac{k(k+1)}{2!} (qe^t)^2 + \frac{k(k+1)(k+2)}{3!} (qe^t)^3 + \dots \right] \\&= (pe^t)^k (1 - qe^t)^{-k}\end{aligned}$$

$$\left[ (1-x)^{-n} = 1 + nx + \frac{n(n+1)}{2!} x^2 + \frac{n(n+1)(n+2)}{3!} x^3 + \dots \right] \text{ provided } |x| < 1$$

$$\therefore M_X(t) = \frac{(pe^t)^k}{(1 - qe^t)^k}, \text{ provided } |qe^t| < 1.$$



$$E(x) = \frac{d}{dt} M_x(t)$$

$$= \frac{(1-qe^t)^k k (pe^t)^{k-1} pe^t - (pe^t)^k k (1-qe^t)^{k-1} \cdot (-qe^t)}{(1-qe^t)^{2k}}$$

$$= \frac{k(1-qe^t)^{k-1} (pe^t)^k [1-qe^t + qe^t]}{(1-qe^t)^{2k}}$$

$$= \frac{k(1-qe^t)^{k-1} (pe^t)^k}{(1-qe^t)^{2k}}$$

$$\text{At } t=0, \mu = \frac{k(1-q)^{k-1} p^k}{(1-q)^{2k}} = \frac{k p^{k-1} p^k}{p^{2k}} = \frac{k p^{2k-1}}{p^{2k}}$$

$$\boxed{\mu = \frac{k}{p}}$$

$$E(x^2) = \frac{d^2}{dt^2} M_x(t) = \frac{d}{dt} \left[ \frac{k(1-qe^t)^{k-1} (pe^t)^k}{(1-qe^t)^{2k}} \right]$$

$$= k \left[ \frac{(1-qe^t)^{2k} \frac{d}{dt} [(1-qe^t)^{k-1} (pe^t)^k] - (1-qe^t)^{k-1} (pe^t)^k \frac{d}{dt} (1-qe^t)^{2k}}{(1-qe^t)^{4k}} \right]$$

$$= k \left[ \frac{(1-qe^t)^{2k} (1-qe^t)^{k-1} k (pe^t)^{k-1} (pe^t) + (pe^t)^k (k-1) (1-qe^t)^{k-2} (-qe^t) + 2k (1-qe^t)^{3k-2} (pe^t)^k qe^t}{(1-qe^t)^{4k}} \right]$$

At  $t=0$ ,

$$E(x^2) = k \left[ \frac{(1-q)^{2k} (1-q)^{k-1} k p^{k-1} p + (k-1) p^k (1-q)^{k-2} (-q) + 2k q p^k (1-q)^{3k-2}}{(1-q)^{4k}} \right]$$

$$= k \left[ \frac{k p^{2k+k-1+k-1+1} + (k-1)(-q) p^{k+k-2} + 2k q p^{k+3k-2}}{p^{4k}} \right]$$

$$= k \left[ \frac{k p^{4k-1} - q(k-1) p^{4k-2} + 2k q p^{4k-2}}{p^{4k}} \right]$$

$$= k \left[ \frac{k p^{4k-1} - q k p^{4k-2} + q p^{4k-2} + 2k q p^{4k-2}}{p^{4k}} \right]$$

$$= k \left[ \frac{k p^{4k-1} + q p^{4k-2} + q k p^{4k-2}}{p^{4k}} \right]$$

$$= k \cdot \frac{p^{4k}}{p^{4k}} \left[ \frac{k}{p} + \frac{q}{p^2} + \frac{qk}{p^2} \right] = k \left[ \frac{k}{p} + \frac{q}{p^2} + \frac{qk}{p^2} \right]$$



$$= k \left[ \frac{pk + q + qk}{p^2} \right]$$

$$= k \left[ \frac{k+q}{p^2} \right]$$

$$= \frac{k}{p^2} [k+q]$$

$$\begin{aligned} \text{Variance} &= \frac{k}{p^2} (k+q) - \frac{k^2}{p^2} = \frac{k^2}{p^2} + \frac{kq}{p^2} - \frac{k^2}{p^2} \\ &= \frac{kq}{p^2} \end{aligned}$$

$$\text{Variance} = \frac{kq}{p^2}$$

## m.g.f. of Geometric Distribution

$$M_X(t) = E[e^{tx}]$$

$$= \sum_{x=1}^{\infty} e^{tx} p q^{x-1}$$

$$= \frac{p}{q} \sum_{x=1}^{\infty} e^{tx} q^x$$

$$= \frac{p}{q} \sum_{x=1}^{\infty} [q e^t]^x$$

$$= \frac{p}{q} \frac{q e^t}{1 - q e^t}, \text{ provided } q e^t < 1.$$

$$= \frac{p e^t}{1 - q e^t} \text{ provided } q e^t < 1.$$

Infinite G.P

$$\sum_{n=1}^{\infty} a^n = \frac{a}{1-r},$$

provided common ratio  $r < 1$

$$M_X(t) = \frac{p e^t}{1 - q e^t} \text{ provided } q e^t < 1$$

### Mean and Variance

At  $t=0$ ,

$$E(X) = \frac{(1 - q e^t) p e^t - p e^t (-q e^t)}{(1 - q e^t)^2}$$

$$= p e^t [1 - q e^t + q e^t] / (1 - q e^t)^2$$



$$E(x) = \frac{pe^t}{(1-qe^t)^2}$$

$$E(x) = \frac{p}{(1-q)^2} = \frac{p}{p^2} = \frac{1}{p} \Rightarrow \boxed{\mu = \frac{1}{p}}$$

$$E(x^2) = \frac{(1-qe^t)^2 pe^t - pe^t \cdot 2(1-qe^t)(-qe^t)}{(1-qe^t)^4}$$

At  $t=0$ ,

$$E(x^2) = \frac{(1-q)^2 p + 2pq(1-q)}{(1-q)^4}$$

$$= \frac{p^3 + 2p^2q}{p^4}$$

$$= \frac{1}{p} + \frac{2q}{p^2}$$

$$\text{Variance} = \frac{1}{p} + \frac{2q}{p^2} - \frac{1}{p^2} = \frac{p + 2q - 1}{p^2} = \frac{1 + q - 1}{p^2}$$

$$\boxed{\text{Variance} = \frac{q}{p^2}}$$

## M.g.f. of Poisson Distribution

$$M_X(t) = E[e^{tx}] = \sum_{x=0}^{\infty} e^{tx} \frac{e^{-\lambda} \lambda^x}{x!}$$

$$= e^{-\lambda} \sum_{x=0}^{\infty} \frac{(e^t \lambda)^x}{x!}$$

$$= e^{-\lambda} e^{e^t \lambda}$$

$$\cancel{= e^{-\lambda(1-e^t)}}$$

$$M_X(t) = e^{-\lambda(1-e^t)}$$

$$\left[ e^x = 1 + x + \frac{x^2}{2!} + \dots \right]$$

## Mean and Variance of Poisson Distribution

$$\begin{aligned} E(X) &= \frac{d}{dt} [M_X(t)] = e^{-\lambda(1-e^t)} (-\lambda)(-e^t) \\ &= (\lambda e^t) e^{-\lambda(1-e^t)} \\ &= \lambda e^{t-\lambda+\lambda e^t} \end{aligned}$$

$$\text{At } t=0, E(X) = \lambda e^{-\lambda+\lambda} = \lambda e^0 = \lambda$$

$$\boxed{\therefore E(X) = \lambda}$$

$$E(X^2) = \lambda \left[ e^t \{-\lambda(-e^t) e^{-\lambda(1-e^t)}\} + e^t e^{-\lambda(1-e^t)} \right]$$

$$\begin{aligned} \text{At } t=0, E(X^2) &= \lambda [1 + \lambda e^{-\lambda(1-1)} + e^{-\lambda(1-1)}] \\ &= \lambda [1 + \lambda + 1] = 2\lambda + \lambda^2 \end{aligned}$$

$$\text{Variance} = 2\lambda + \lambda^2 - \lambda^2 =$$



$$E(X^2) = \frac{d^2}{dt^2} [M_X(t)]$$

$$= \frac{d}{dt} [(\lambda e^t) e^{-\lambda(1-e^t)}]$$

$$= \lambda \left[ e^t e^{-\lambda(1-e^t)} + e^t e^{-\lambda(1-e^t)} (+\lambda e^t) \right]$$

At  $t=0$

$$E(X^2) = \lambda [e^{-\lambda(1-1)} + e^0 e^{-\lambda(1-1)} \lambda]$$

$$= \lambda [1 + \lambda] = \lambda + \lambda^2$$

$$\text{Variance} = \lambda + \lambda^2 - \lambda^2 = \lambda$$

$$\text{Variance} = \lambda$$