

Random Variable :-

DRV

cRV

continuous R.V

Prf: (1) $P(x) \geq 0$

Df: (1) $P(x) \geq 0$

(2) $\sum P(x) = 1$

(2) $\int f(x) dx = 1$

R.V is a real valued fm which assign a real number to each sample point in the sample space.

Eg: Tossing a fair coin twice then,

Sample space $S = \{HHH, HHT, HTH, THH, HTT, THT, TTH, TTT\}$

X : no. of heads

$$x(S_1) = 3$$

$$x(S_2) = x(S_3) = x(S_4) = 2$$

$$x(S_5) = x(S_6) = x(S_7) = 1$$

$$x(S_8) = 0$$

i) RV: If R.V which takes finite or at most countable no. of variables values is called discrete R.V.

Eg: (1) no. of heads obtained when two coins are tossed.

(2) no. of defective items in a bag.

Probability distribution

x (no. of head)	0	1	2	3
P	$\frac{1}{8}$	$\frac{3}{8}$	$\frac{3}{8}$	$\frac{1}{8}$

100 bad oranges are mixed with 16 good oranges. Find probability distribution of no. of bad oranges in a draw of 100 oranges.

Soln:

x (no. of bad orange)	0	1	2	3	4
P	$\frac{16}{100}C_0$	$\frac{16}{100}C_1$	$\frac{16}{100}C_2$	$\frac{16}{100}C_3$	$\frac{16}{100}C_4$

16 good	$\frac{16}{100}C_0$
4 bad	$\frac{16}{100}C_1$

Probability mass fm (P.m.f):

Let X be a R.V D.R.V. s.t $P(X=x) = p_i$, then p_i is said to be probability mass function (P.m.f) if it satisfy the following condition:

$$(1) p_i \geq 0$$

$$(2) \sum p_i = 1$$

Ex - $S = \{HHH, HHT, HTH, THH, HTT, THT, TTH, TTT\}$

Probability dist.

x (no. of heads)	0	1	2	3
P	$\frac{1}{8}$	$\frac{3}{8}$	$\frac{3}{8}$	$\frac{1}{8}$

Distribution function -

Let X be a d.R.V., then its discrete distribution f.m or cumulative distribution f.m (c.d.f) is defined as
 $f(x) = \sum p_i = P(X \leq x)$

$$f(x) = \begin{cases} 0 & x \leq 0 \\ \frac{1}{8} & x \leq 1 \\ \frac{7}{8} & x \leq 2 \\ 1 & x \leq 3 \end{cases}$$

Q) Given the following probability distribution -

x	0	1	2	3	4	5	6	7
$p(x)$	0	k	$2k$	$3k$	k^2	$2k^2$	$7k^2 + k$	

Find ① K ② $P(X < 6)$, $P(X \geq 6)$, $P(0 < X < 5)$

③ Distribution f.m

④ If $P(X \leq c) > 1$, find value of c

⑤ $P(1.5 < X < 4.5)$

Ques: If $p(x) = P_{m,n}$

$$\sum p(x) = 1$$

$$0 + k + 2k + 2k + 3k + k^2 + 2k^2 + 7k^2 + K = 1$$

$$10k^2 + 9k - 1 = 0 \Rightarrow 10k^2 + 10k - k - 1 = 0$$

$$\Rightarrow 10k(k+1) - 1(k+1) = 0$$

$$k = -1 \quad | \quad k = \frac{1}{10}$$

not possible
can't
be -ve.

So,

x	0	1	2	3	4	5	6	7
$p(x)$	0	0.1	0.2	0.2	0.3	0.01	0.02	0.17

$$\textcircled{2} \quad P(X \geq 6) = P(X < 6)$$

$$= p(0) + p(1) + p(2) + p(3) + p(4) + p(5) \quad \text{or} \quad 1 - P(X \geq 6)$$

$$1 - [p(6) + p(7)] = 1 - [0.09] \\ = 0.81$$

$$P(X \geq 6)$$

$$p(6) + p(7) = 0.09$$

$$P(0 < X < 5)$$

$$= p(1) + p(2) + p(3) + p(4)$$

$$= 0.1 + 0.2 + 0.2 + 0.3 = 0.8$$

③

0	$x \leq 0$
0.1	$x \leq 1$
0.3	$x \leq 2$
0.5	$x \leq 3$
0.8	$x \leq 4$
0.81	$x \leq 5$
0.83	$x \leq 6$
1	$x \leq 7$

① for $x \geq 4$ $P(x) = 0.8 > \frac{1}{2}$
 $\therefore c = 1$

⑤ $P(1.5 < x < 4.5)$ \rightarrow
 $x \geq 2$

$$P(A) = \begin{cases} P(A \cap B) & P(A \cap B) \\ P(B) & P(B) \end{cases}$$

So,
 $P(1.5 < x < 4.5) \cap P(x \geq 2) = P(2, 3, 4) \cap P(3, 4, 5, 6, 7)$
 $P(x \geq 2) = P(3, 4, 5, 6, 7)$

$$= \frac{P(3, 4)}{P(x \geq 2)} = \frac{P(3) \cdot P(4)}{1 - P(x \leq 2)} = \frac{0.5 \cdot 0.5}{1 - 0.3} = \frac{0.5}{0.7} = \frac{5}{7}$$

Continuous Random Variable :-

* A R.V which can take infinite no. of values in an interval is known as CRV.

- Eg: i) The weight of a group of individuals.
 ii) Height of a group of individuals.
 iii) Price of house

Probability density function (P.d.f) -

- A $f(x)$ is p.d.f if -
 ① $f(x) \geq 0$ $-\infty < x < \infty$

② $\int_{-\infty}^{\infty} f(x) dx = 1$

It is also known as density function.

Eg: If X is a CRV with the following p.d.f

$$f(x) = \begin{cases} \alpha(2x-x^2) & 0 < x < 2 \\ 0 & \text{otherwise} \end{cases}$$

find ① α ② $P(x > 1)$

So: By defn of p.d.f $\int_{-\infty}^{\infty} f(x) dx = 1$

$$\int_{-\infty}^{\infty} f(x) dx + \int_{-1}^0 f(x) dx + \int_0^2 f(x) dx + \int_2^{\infty} f(x) dx = 1$$

$$\Rightarrow 0 + \int_{-1}^0 f(x) dx + \int_0^2 f(x) dx = 1$$

$$\Rightarrow \int_0^2 \alpha(2x-x^2) dx = 1 \Rightarrow \alpha \left[x^2 - \frac{x^3}{3} \right]_0^2 = 1$$

$$\Rightarrow \alpha \left[(4-8) - \frac{8}{3} \right] = 1$$

$$\Rightarrow \alpha \left(\frac{4}{3} \right) = 1 \quad P(x > 2) = \int_2^{\infty} f(x) dx \\ \Rightarrow \alpha = \frac{3}{4} \quad = \int_2^{\infty} \alpha(2x-x^2) dx = \int_2^{\infty} f(x) dx$$

$$= \frac{3}{4} \left[x^2 - \frac{x^3}{3} \right]_2^{\infty} = \frac{3}{4} \left[(3-1) - \frac{(3-1)^3}{3} \right] = \frac{3}{4} \left[3 - \frac{8}{3} \right] = \frac{3}{4} \left[\frac{1}{3} \right] = \frac{3}{12} = \frac{1}{4}$$

Q) A R.V X has density $f(x)$

$$f(x) = \begin{cases} kx^2 & -3 < x < 3 \\ 0 & \text{otherwise} \end{cases}$$

Find K , $P(1 \leq x \leq 2)$, $P(x \leq 2)$, $P(x > 1)$

Soln: By the defⁿ of pdf -

$$\int_{-\infty}^{\infty} f(x) \cdot dx = 1$$

$$\Rightarrow \int_{-3}^3 kx^2 \cdot dx = 1 \Rightarrow kx^3 \Big|_{-3}^3 = 1$$

$$\Rightarrow K [9 + 9] = 18K \cdot 1 \Rightarrow K = 1/18$$

$P(1 \leq x \leq 2)$

$$\frac{1}{18} \int_1^2 x^2 \cdot dx = \frac{1}{18} \left[x^3 \right]_1^2 = \frac{1}{18} \left[x^3 \right]_1^2 = \frac{1}{18} \left(8 - \frac{1}{3} \right) = \frac{1}{18} \left(\frac{24}{3} - \frac{1}{3} \right) = \frac{1}{18} \left(\frac{23}{3} \right) = \frac{13}{54}$$

$P(x \leq 2)$

$$\int_{-\infty}^2 f(x) \cdot dx = \int_{-\infty}^2 kx^2 \cdot dx = \frac{1}{18} \left[x^3 \right]_{-\infty}^2 = \frac{1}{18} \left[x^3 \right]_{-3}^2 = \frac{1}{18} (8 - (-27)) = \frac{1}{18} (35) = \frac{35}{18}$$

$$= \frac{1}{18} \left(\frac{8 + 27}{3} \right) = \frac{1}{18} \left(\frac{35}{3} \right) = \frac{35}{54}$$

$P(x > 1)$

$$\int_1^3 f(x) \cdot dx = \frac{1}{18} \int_1^3 x^2 \cdot dx = \frac{1}{18} \left[x^3 \right]_1^3 = \frac{1}{18} \left[x^3 \right]_1^3 = \frac{1}{18} \left(\frac{27 - 1}{3} \right) = \frac{1}{18} \left(\frac{26}{3} \right) = \frac{13}{27}$$

$$= \frac{1}{18} \left(\frac{26}{3} \right) = \frac{13}{27}$$

• Cumulative distribution function (c.d.f)

continuous

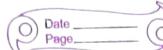
Let X be continuous random variable having pdt $f(x)$
then $F_x(x)$ will be a continuous distribution function
of x if

$$F_x(x) = P(X \leq x) = \int_{-\infty}^x f(x) \cdot dx$$

this is also known as cumulative distribution function.

1.00
2.25
1.75

1.00
0.875
0.125



Note:- Relation b/w distribution function and density function
$$dF(x) = f(x) dx$$

Eg: The probability density function of random variable, x is

$$f(x) = \begin{cases} n & 0 \leq x < 1 \\ 2-x & 1 < x < 2 \\ 0 & otherwise \end{cases}$$

① Find $P(x \geq 1.5)$

② Find cumulative distribution function

Soln:
$$P(x \geq 1.5) = \int_{1.5}^{\infty} f(x) dx = \int_{1.5}^2 (2-x) dx$$

$$= 2x - \frac{x^2}{2} \Big|_{1.5}^2 = 2(0.5) - \frac{1}{2}(4-2.25) = 1 - \frac{1}{2}(1.75)$$

$$= 1 - \frac{175}{200} = 1 - 0.875 = 0.125$$

② $x \leq 0$
 $P(x \leq x) = \int_{-\infty}^x f(x) dx = 0$

③ $0 \leq x < 1$
~~If~~ $F(x) = \int_{-\infty}^0 f(x) dx + \int_0^x f(x) dx$

$$f(x) = 0 + \int_{-\infty}^x (2-x) dx = 0 + \frac{x^2}{2}$$

$1 < x < 2$

$$\int_{-\infty}^x (2-x) dx = \int_1^x (2-x) dx = \int_1^2 (2-x) dx = 2\left(\frac{1}{2}\right) \cdot \left(2 - \frac{1}{2}\right) = 2 - \frac{3}{2} = \frac{1}{2}$$

$$= \frac{x^2}{2} \Big|_0^1 = \left(2x - \frac{x^2}{2}\right) \Big|_0^1 = \frac{1}{2} + \left(2 \cdot \frac{1}{2}\right) - \left(2 \cdot \frac{1}{2}\right)$$

$$= \frac{1}{2} - 2 + \frac{1}{2} + 2x - x^2$$

$x \geq 2$

$$f(x) = \int_{-\infty}^0 f(x) dx + \int_0^1 f(x) dx + \int_1^2 f(x) dx + \int_2^{\infty} f(x) dx$$

$$= 0 + \frac{x^2}{2} \Big|_0^1 + \left(2x - \frac{x^2}{2}\right) \Big|_1^2 + 0$$

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

$$= \frac{1}{2} + \left[2 - \frac{3}{2} \right] = \frac{1}{2} + \frac{1}{2} = 1$$



$$f_x(x) = \begin{cases} 0 & -\infty < x < 0 \\ x^2/2 & 0 < x < 1 \\ -1 + 2x - x^2/2 & 1 \leq x \leq 2 \end{cases}$$

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Mathematical Expectation -

Let x be any R.V & $\phi(x)$ be any function of x . Then expectation of $\phi(x)$ is denoted by $E(\phi(x))$ & is defined by

$$E(\phi(x))$$

$$\sum_{x} \phi(x) \cdot p(x) \quad \text{C.R.V}$$

$\int_{-\infty}^{\infty} \phi(x) \cdot f(x) \cdot dx$

↓
pmf

$$\text{if } \phi(x) = x, \quad E(x) = \sum_{x} x \cdot p(x)$$

$$E(x) \quad \text{D.R.V}$$

$$E(x) = \int_{-\infty}^{\infty} x \cdot f(x) \cdot dx$$

$$\text{Variance} = E(x^2) - (E(x))^2$$

$$\text{Proof: } E(x - \bar{x})^2 = E(x^2 - 2x\bar{x} + \bar{x}^2)$$

$$= E(x^2) - 2\bar{x}E(x) + \bar{x}^2$$

$$= E(x^2) - 2\bar{x}^2 + \bar{x}^2$$

$$= E(x^2) - \bar{x}^2$$

$$= E(x^2) - (E(x))^2$$

Ans: \rightarrow

Succ:	P	No. of x
Unsucc:	$1-P$	$n-m$
$\therefore C_m P^m (1-P)^{n-m}$		$n = total$

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Q) Find mean and variance of the probability distribution given by the following table -

x	1	2	3	4	5
$P(x)$	0.2	0.35	0.25	0.15	0.05

$$E(x) = \sum x \cdot P(x) = 0.2 + 0.7 + 0.75 + 0.6 + 0.25$$

mean = 2.5

$$E(x^2) = \sum x^2 \cdot P(x) = 0.2 + 1 \cdot 1 + 2.25 + 2 \cdot 4 + 3.25 = 7.5$$

$$\text{Var}(x) = \sigma^2 x = E(x^2) - (E(x))^2$$

$$= 7.5 - (2.5)^2 = 7.5 - 6.25$$

$$= 1.25$$

Q) Thirteen cards are drawn simultaneously from a pack of 52 cards. If ace count 1 and face card 10 and other according to their denomination. find the expectation of total score in 13 cards.

Soln:

x	1	2	3	4	5	6	7	8	9	10	10	10	10
$P(x)$	4/52	1/52	1/52	1/52	1/52	"	"	"	"	"	"	"	"

↓
1/13

$$E(x) = \sum x \cdot P(x) = \frac{1}{13} [1+2+3+4+5+6+7+8+9+10+10+10+10]$$

$$= 85/13$$

Q). A continuous R.V X has density function given by

$$f(x) = \begin{cases} 2e^{-2x} & x > 0 \\ 0 & \text{otherwise} \end{cases}$$

Find expected value & variance of X

Sol:

$$E(x) = \int_{-\infty}^{\infty} x \cdot f(x) \cdot dx$$

$x > 0$

$$\int_{0}^{\infty} x \cdot 2e^{-2x} \cdot dx = 2 \int_{0}^{\infty} x \cdot e^{-2x} \cdot dx$$

Variance

$$E(x^2) = \int_{-\infty}^{\infty} x^2 \cdot f(x) \cdot dx = \int_{0}^{\infty} x^2 \cdot 2e^{-2x} \cdot dx$$

$$\text{Var.} = E(x^2) - (E(x))^2$$

Binomial Distribution:

- ① All the trials are independent
- ② Number (n) of trial is fixed
- ③ The probability of success is same for each trial.

$P(n) = {}^n C_x p^x (1-p)^{n-x}$
 $n \rightarrow \text{no. of trials.}$
 $x \rightarrow \text{successful trial.}$
 $p \rightarrow \text{probability of successful event.}$

Show that $P(n)$ is pmf.

→

for pmf: ① $P(n) \geq 0$

$$\textcircled{2} \sum P(n) = 1$$

$$\sum P(n) = \sum_{x=0}^n {}^n C_x p^x (1-p)^{n-x}$$

$$= (1-p)^n + {}^n C_1 p^1 (1-p)^{n-1} + \dots + {}^n C_n p^n$$

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

Hence, proved.

Find moment generating function of Binomial distribution

$$M.G.F = M_X(t) = E(e^{xt})$$

$$= \sum_{x=0}^n e^{xt} P(n) = \sum_{x=0}^n e^{xt} {}^n C_x p^x (1-p)^{n-x}$$

$$= \sum_{x=0}^n {}^n C_x (p e^t)^x \cdot (1-p)^{n-x}$$

$$\Downarrow$$

$$(p e^t + 1 - p)^n$$

So,

$$M_X(t) = (p e^t + (1-p))^n$$

Characteristic function -

$$\phi_Y(z) = E(e^{izt})$$

$$= [P.e^{izt} + (1-p)]^n$$

Probability generating function -

$$Z_X(z) = E(z^x)$$

$$= \sum_{x=0}^n z^x \cdot P(x) = \sum_{x=0}^n z^x \cdot {}^n C_x p^x (1-p)^{n-x}$$

$$= \sum_{x=0}^n {}^n C_x (zp)^x \cdot (1-p)^{n-x}$$

$$= (zp + (1-p))^n$$

Mean & Variance of Binomial distribution -

$$\boxed{\text{Mean} = E(x) = np}$$

$$\begin{aligned} \text{Variance} &= \sigma^2 = E(x^2) - (E(x))^2 \\ &= (np)^2 + npq - (np)^2 \\ &= npq \end{aligned}$$

$$\text{here, } q = 1-p$$

i.e.

$$\boxed{\sigma^2 = np(1-p)}$$

Q) The probability that man aged 60 will live upto 70 is 0.65 out of 10 men, now aged 60 find probability -

(1) At least 7 will live upto 70

(2) Exactly 9 will live upto 70

(3) At most 9 will live upto 70

Soln: Q) At least 7 will live upto 70 $n = 10 \quad p = 0.65 \quad 1-p = 0.35$

$$P(x) = {}^{10} C_x 0.65^x (0.35)^{10-x}$$

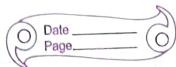
$$(1) P\{x \geq 7\} = 1 - P\{x \leq 6\}$$

$$\begin{aligned} P\{x \geq 7\} &= P(7) + P(8) + P(9) + P(10) \\ &= {}^{10} C_7 (0.65)^7 (0.35)^3 + {}^{10} C_8 (0.65)^8 (0.35)^2 + {}^{10} C_9 (0.65)^9 (0.35)^1 \\ &\quad + {}^{10} C_{10} (0.65)^{10} \end{aligned}$$

$$= 0.5139$$

$$(2) P\{x = 9\} = {}^{10} C_9 (0.65)^9 (0.35)^1 = 0.0725$$

$$\begin{aligned} (3) P\{x \leq 9\} &= 1 - P\{x \geq 10\} \\ &= 1 - P\{x = 10\} \\ &= 1 - {}^{10} C_{10} (0.65)^{10} (0.35)^0 = 0.9865 \end{aligned}$$



- Q) Out of 800 families with 5 children each. How many families would be expected to have (1) 3 boys (2) 5 girls.
 (3) either 2 or 3 boys (4) at least 2 girls.

Solⁿ: N=800 N=800, n=5

$$P(\text{Boys}) = \frac{1}{2}$$

$$1-p = \frac{1}{2} \quad 1-\frac{1}{2} = \frac{1}{2}$$

↓
Probab. of girls

$$P(n) = {}^m C_n p^n (1-p)^{n-m}$$

$${}^5 C_3 \left(\frac{1}{2}\right)^3 \left(\frac{1}{2}\right)^{5-3} = {}^5 C_3 \left(\frac{1}{2}\right)^5 = {}^5 C_3 \frac{1}{32}$$

$$\textcircled{1} \quad P(3) = {}^5 C_3 \frac{1}{32} = \frac{5 \times 4}{2} \left(\frac{1}{2}\right)^5 = \frac{10}{32} \quad \begin{matrix} \text{No. of fam. expected} \\ 10 \times 25 \\ = 250 \end{matrix}$$

$$\textcircled{2} \quad \text{Probab. of 5 girls} = \text{Probab. of 0 boys} \\ : P(0)$$

$$\text{i.e. } {}^5 C_0 \left(\frac{1}{2}\right)^5 = \frac{1}{32}$$

$$\text{No. of families with 5 cr} = \frac{1}{32} \times 800$$

$$\textcircled{3} \quad P(2) + P(3) \\ = {}^5 C_2 \left(\frac{1}{2}\right)^2 \left(\frac{1}{2}\right)^3 + {}^5 C_3 \left(\frac{1}{2}\right)^3 \left(\frac{1}{2}\right)^2 = \frac{10}{32} + \frac{10}{32} = \frac{10}{16} = \frac{5}{8}$$

$$\text{No. of families} = 800 \left(\frac{5}{8}\right) = 500$$

$$\textcircled{4} \quad P(X > 2) = 1 - P(X < 2) \\ = 1 - P(X=1) \\ = 1 - {}^5 C_1 \frac{1}{32} = 1 - \frac{5}{32} = \frac{27}{32}$$

Q) Probab. of at least 2 girls: 2G 3G 4G 5G

3B 2B 1B 0B

J

i.e we have to find

$$P(2) + P(3) + P(4) + P(5)$$

OR

$$1 - [P(0) + P(1)]$$

$$P(0) = {}^5 C_0 \cdot \frac{1}{32} = \frac{1}{32}$$

$$P(1) = {}^5 C_1 \frac{1}{32} = \frac{5}{32}$$

so,

$$1 - \left(\frac{6}{32}\right) = \frac{26}{32}$$

$$\text{No. of families} = \frac{26}{32} (800)$$

Q) 4 coins are tossed 100 times and following were obtained. Fit a binomial distribution for data & calculate theoretical frequency.

No. of Head(s)	freq.(n)	xf
0	5	0
1	29	29
2	36	72
3	25	75
4	5	20

$$\sum xf = 100 \quad \sum nxf = 196$$

$$np = \bar{x} = \frac{\sum xf}{\sum f} = \frac{1.96}{100} = 1.96$$

$$AP = 1.96$$

$$P = \frac{1.96}{4} = 0.49$$

$$1-p = 0.51$$

$$P(n) = {}^n C_x p^x (1-p)^{n-x}$$

$$= {}^n C_n p^x (0.49)^x (0.51)^{n-x}$$

Total freq = 100

100 P(x)

$$P(0) = {}^0 C_0 (0.49)^0 (0.51)^4 = 0.0676$$

$$67.6 \approx 6.76 \approx 6$$

$$P(1) = {}^1 C_1 (0.49)^1 (0.51)^3 = 0.2511$$

$$25.11 \approx 26$$

$$P(2) = {}^2 C_2 (0.49)^2 (0.51)^2 = 0.3717$$

$$37.17 \approx 37$$

$$P(3) = {}^3 C_3 (0.49)^3 (0.51)^1 = 0.2400$$

$$24.00$$

$$P(4) = {}^4 C_4 (0.49)^4 (0.51)^0 = 0.05765$$

$$5.76 \approx 6$$

$$\vdots$$

Poisson Distribution:-

A discrete P.V X which has the following probability mass function $P(x) = \frac{\lambda^x e^{-\lambda}}{x!}$, $x = 0, 1, 2, \dots$ is called

poisson variable & its distribution is called poisson distribution.

Q) Prove that poisson distribution is a limiting case of Binomial distribution under following condition-

$$① n \rightarrow \infty$$

$$② p \rightarrow 0$$

$$③ np = \lambda \text{ (finite)}$$

Proof: We know that, $P(x) = {}^n C_x p^x (1-p)^{n-x}$

$$\lim_{n \rightarrow \infty} P(x)$$

$$n \rightarrow \infty$$

$$= \lim_{n \rightarrow \infty} {}^n C_x p^x (1-p)^{n-x} = \lim_{n \rightarrow \infty} \frac{n!}{x!(n-x)!} p^x (1-p)^{n-x}$$

$$= \lim_{n \rightarrow \infty} \frac{n(n-1)\dots(n-(x-1))(n-x)!}{x!(n-x)!} \left(\frac{\lambda}{n}\right)^x \left(\frac{1-\lambda}{n}\right)^{n-x}$$

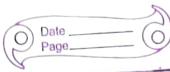
[From ③, $np = \lambda$
 $p = \lambda/n$]

$$= \lim_{n \rightarrow \infty} \frac{n^x}{x!} \left[\left(1 - \frac{1}{n}\right) \dots \left(1 - \frac{(x-1)}{n}\right) \right] \frac{\lambda^x}{x!} \left(\frac{1-\lambda}{n}\right)^{n-x}$$

$$= \lim_{n \rightarrow \infty} \frac{\lambda^x}{x!} \left(\frac{1-\lambda}{n}\right)^n \left(\frac{1-\lambda}{n}\right)^{-x}$$

$$= \frac{\lambda^x}{x!} \lim_{n \rightarrow \infty} \left(\frac{1-\lambda}{n}\right)^n \lim_{n \rightarrow \infty} \left(\frac{1-\lambda}{n}\right)^{-x}$$

$$= \frac{\lambda^x e^{-\lambda}}{x!} = \frac{\lambda^x e^{-\lambda}}{x!}$$



Mean & Variance of Poisson Distribution -

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

$$\text{mean} = E(x) = \sum_n x \cdot P(x) \cdot n! \\ = \sum_n x \cdot \frac{\lambda^x e^{-\lambda}}{x!}$$

Normal Distribution

A continuous R.V which has the following pdf

$$f(x) = \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{1}{2} \left(\frac{x-\mu}{\sigma}\right)^2} \quad -\infty < x < \infty$$

$$-\infty < \mu < \infty$$

$$\sigma > 0$$

is called normal variable & its distribution is called normal distribution & is denoted by $X \sim N(\mu, \sigma)$

Mean & Variance of normal distribution -

$$f(x) = \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{1}{2} \left(\frac{x-\mu}{\sigma}\right)^2}$$

$$E(x) = \int_{-\infty}^{\infty} x \cdot f(x) dx = \int_{-\infty}^{\infty} x \cdot \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{1}{2} \left(\frac{x-\mu}{\sigma}\right)^2} dx$$

$$\boxed{E(x) = \mu}$$

$$E(x^2) = \int_{-\infty}^{\infty} x^2 \cdot f(x) dx = \boxed{\mu^2 + \sigma^2}$$

$$\Rightarrow \text{Variance} = E(x^2) - (E(x))^2 \\ = \mu^2 + \sigma^2 - \mu^2 \\ = \sigma^2$$

Moment generating function of Normal Distribution 8

$$f(x) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

$-\infty < x < \infty$

$$M_x(t) = E(e^{xt}) = \int_{-\infty}^{\infty} e^{xt} f(x) dx$$

$$= \int_{-\infty}^{\infty} e^{xt} \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx$$

$$= \frac{1}{\sigma\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{xt} \frac{1}{\sigma} e^{-\frac{(x-\mu)^2}{\sigma^2}} dx$$

$$\text{let } \frac{x-\mu}{\sigma} = z$$

$$\text{so, } dx = dz$$

$$= \frac{1}{\sigma\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{(\mu+\sigma z)t} e^{-\frac{z^2}{2}} dz$$

$$= \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{\mu t} e^{\sigma z t} e^{-\frac{z^2}{2}} dz$$

$$= \frac{e^{\mu t}}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-\frac{z^2}{2}} e^{\sigma^2 z^2/2} dz$$

$$= \frac{e^{\mu t + \sigma^2 t^2/2}}{\sqrt{2\pi}} \int_0^{\infty} e^{-z^2/2} (1 - e^{-2\mu z}) dz$$

$$= \frac{2e^{\mu t + \sigma^2 t^2/2}}{\sqrt{2\pi}} \int_0^{\infty} e^{-z^2/2} dz$$

$$M_x(t) = \frac{2e^{\mu t + \sigma^2 t^2/2}}{\sqrt{2\pi}} \int_0^{\infty} e^{-\theta} d\theta$$

$$= \frac{2e^{\mu t + \sigma^2 t^2/2}}{2\sqrt{\pi}} \int_0^{\infty} \theta^{-1/2} e^{-\theta} d\theta$$

$$M_x(t) = e^{\mu t + \sigma^2 t^2/2}$$

MGF of standard normal variable z .

$$z = \frac{x-\mu}{\sigma}$$

$$M_z(t) = E(e^{zt}) = E(e^{(z-\mu)+\mu})$$

$$= E\left(e^{zt/\sigma} \cdot e^{-\mu t/\sigma}\right) = e^{-\mu t/\sigma} E\left(e^{zt/\sigma}\right)$$

$$= e^{-\mu t/\sigma} M_x(t/\sigma) = e^{-\mu t/\sigma} \cdot e$$

$$M_z(t) = e^{-t^2/2}$$

Mean & Variance by MGF -

$$Mx(t) = e^{\mu t + \sigma^2 t^2/2}$$

$$E(x) = \frac{d}{dt} (Mx(t)) \Big|_{t=0}$$

$$= \frac{d}{dt} (e^{\mu t + \sigma^2 t^2/2}) \Big|_{t=0}$$

$$= e^{\mu t + \sigma^2 t^2/2} (\mu + \sigma^2 t) \Big|_{t=0}$$

$$= e^{\mu t} (\mu)$$

$$[E(x) = \mu]$$

$$E(x^2) = \frac{d^2}{dt^2} (Mx(t)) \Big|_{t=0}$$

$$= \frac{d}{dt} (e^{\mu t + \sigma^2 t^2/2} (\mu + \sigma^2 t)) \Big|_{t=0}$$

$$= e^{\mu t + \sigma^2 t^2/2} (\mu + \sigma^2 t)^2 + \mu e^{\mu t + \sigma^2 t^2/2} (2\sigma^2)$$

$$= \mu^2 + \mu \sigma^2$$

$$\text{Variance} = E(x^2) - (E(x))^2$$

$$= \mu^2 + \mu \sigma^2 - \mu^2 = \mu \sigma^2$$

Moments :-

Moment about origin

Moment about mean

Moment about any other point

$$\mu_1' = E(x - 0)^\alpha$$

$$\mu_2 = E(x - \bar{x})^\alpha$$

$$\mu_3''' = E(x - A)^\alpha$$

$$\mu_1' = E(x)^\alpha$$

$$\mu_2'' = E(x^2)^\alpha$$

$$\mu_3''' = E(x^3)^\alpha$$

$$\mu_4'''' = E(x^4)^\alpha$$

$$\mu_5''' = E(x^5)^\alpha$$

$$\mu_6'''' = E(x^6)^\alpha$$

$$\mu_7''' = E(x^7)^\alpha$$

$$\mu_8'''' = E(x^8)^\alpha$$

$$\mu_9''' = E(x^9)^\alpha$$

$$\mu_1' = E(x - \bar{x})^\alpha$$

$$= E(x^2 - 2x\bar{x} + \bar{x}^2)$$

$$= E(x^2) - (E(x))^\alpha$$

$$= \text{variance}$$

$$\boxed{\mu_2 = \mu_1' - 3\mu_1\mu_1' + 2(\mu_1')^2}$$