

PROBABILITY

① $P = \frac{\text{Required}}{\text{Total}}$

② $0 \leq P(A) \leq 1$
 Null/Impossible event \rightarrow Sure event

③ $P(\bar{A}) = 1 - P(A)$ ($\bar{A}/A'/A^c$)

④ $P(A \cup B) = P(A) + P(B) - P(A \cap B)$

⑤ $P(A \cup B \cup C) = P(A) + P(B) + P(C) - P(A \cap B) - P(B \cap C) - P(A \cap C) + P(A \cap B \cap C)$

⑥ $P(\overline{A \cap B}) = P(\bar{A} \cap \bar{B}) = 1 - P(A \cup B)$

⑦ $P(\overline{A \cap B}) = P(\bar{A} \cup \bar{B}) = 1 - P(A \cap B)$

⑧ $P(\text{Only } A) = P(A \cap \bar{B}) = P(A) - P(A \cap B)$

* COIN Toss (Independent event)
 'n' coin is tossed $\Rightarrow n(S) = 2^n$

⑨ $P(\text{both } A \text{ \& } B) = P(A \cap B)$

⑩ $P(\text{at least one}) = P(\text{either } A \text{ or } B) = 1 - P(\text{none}) = P(A \cup B)$

⑪ $P(\text{neither } A \text{ nor } B) = P(\bar{A} \cap \bar{B}) = P(\overline{A \cup B})$

⑫ $P(\text{exactly one}) = P(A \cap \bar{B}) + P(\bar{A} \cap B)$

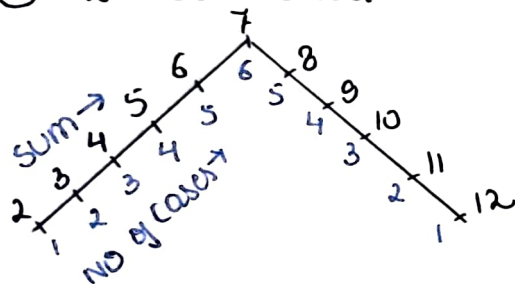
⑬ Mutually Exclusive events:
 can't occur together
 $P(A \cap B) = 0$; $A \cap B = \Phi$

⑭ Independent Events: occurrence of 1 doesn't affect other
 $P(A \cap B) = P(A) \cdot P(B)$

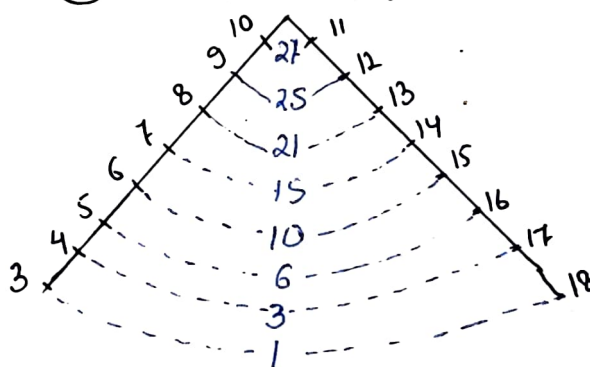
⑮ Dependent Events:
 $P(A \cap B) = P(A) \cdot P(B/A) = P(B) \cdot P(A/B)$

* DICE THROW (Independent event)
 'n' dice throw $\Rightarrow n(S) = 6^n$

⑯ 2 Dice rolled

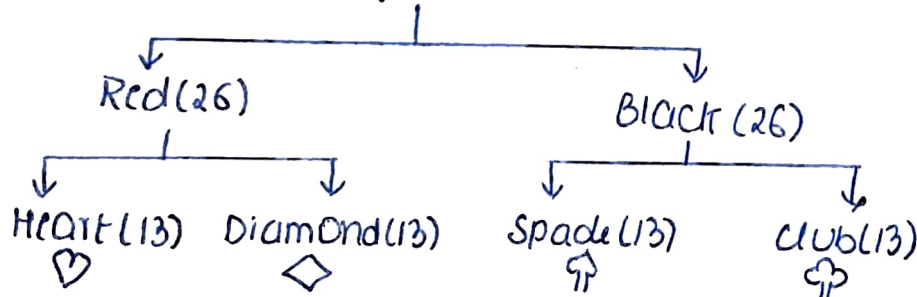


⑰ 3 Dice rolled



⑱ If no. of coin > 3 , no. of dice $> 2 \Rightarrow$ use Binomial dist.

Pack of Card (52)



⑲ Each suit contains

(2, 3, 4, 5, 6, 7, 8, 9, 10)

Number cards = 9

(J, K, Q, A) \rightarrow Honour cards = 4
 Face cards = 3

⑳ If A & B are independent events

- $\bar{A} \text{ \& } B$ " " "
- $A \text{ \& } \bar{B}$ " " "
- $\bar{A} \text{ \& } \bar{B}$ " " "

㉑ $P(A/B) = \frac{P(A \cap B)}{P(B)}$

㉒ $P(A/B) + P(\bar{A}/B) = 1$

$P(A/\bar{B}) + P(\bar{A}/\bar{B}) = 1$

* $P(A) = P(A/x)P(x) + P(A/y)P(y)$
 * If x is average, there must be at least 1 data point $\geq x$ & $\leq x$.

㉓ By default \rightarrow without Replacement.

- (24) A, B, C are pair wise independent events
 $P(A \cap B) = P(A) \cdot P(B)$
 $P(B \cap C) = P(B) \cdot P(C)$
 $P(A \cap C) = P(A) \cdot P(C)$
 \hookrightarrow All possible pairs
- (25) A, B, C are mutually independent events
 $P(A \cap B) = P(A) \cdot P(B)$
 $P(B \cap C) = P(B) \cdot P(C)$
 $P(C \cap A) = P(C) \cdot P(A)$
 $P(A \cap B \cap C) = P(A) \cdot P(B) \cdot P(C)$
 \hookrightarrow All possible combinations of A, B & C
- (26) Total no. of conditions for mutually independence of n events = $2^n - 1 - n$
- (27) Mutually independent events are pair wise independent but not vice-versa.

PDF: Prob. density / Mass fuⁿ
 CDF: Cumulative distribution fuⁿ

RANDOM VARIABLE

\hookrightarrow value of x is discrete set of nos.
 * Discrete Random variable (Σ)
 $P(x)$ is PDF.

PDF: $\sum_{-\infty}^{\infty} P(x) = 1$

CDF: $F(x) = \sum_{-\infty}^x P(x)$

\hookrightarrow value of x is given in range
 * Continuous Random variable
 $f(x)$ is PDF

PDF: $\int_{-\infty}^{\infty} f(x) dx = 1$
 Area under curve = 1

CDF: $F(x) = \int_{-\infty}^x f(x) dx$

(1) CDF: $F(x) = P(X \leq x)$

$F(-\infty) = 0$

$F(\infty) = 1$

(2) $P(X=2) = F(X=2) - F(X=1)$
 (in discrete RV)

(3) CDF is continuous fuⁿ
 (for cont. RV)

CDF is discontinuous fuⁿ
 (for discrete RV)

(4) $0 \leq F(x) \leq 1$

(5) CDF \leftrightarrow PDF; both τ inter convertible.

(4) PDF: $f(x) = \frac{d}{dx} F(x)$

(5) $P(x_1 < X < x_2) = \int_{x_1}^{x_2} f(x) dx$
 (for cont RV)

(6) In cont. RV

$P(a < x < b) = P(a \leq x < b)$
 $= P(a < x \leq b) = P(a \leq x \leq b)$

(7) In discrete RV all are different

$P(0 < x < 2) = P(1)$

$P(0 < x \leq 2) = P(0) + P(2)$

2D Random variables

• Discrete RV

$\sum_{y=x} \sum P(x, y) = 1$

• Continuous RV

$\iint_{yx} f(x, y) dx dy = 1$

Marginal PDF

• eliminate others
 Given $f(x, y)$

• $f(x) = \sum_y f(x, y)$
 $y =$ (in discrete)
 Range of x must be constant

• $f(x) = \int_y f(x, y) dy$
 $y =$ (cont.)

• X & Y are independent

$f(x, y) = f(x) \cdot f(y)$

Mean/Expectation/Average

(1) $E(X) = \sum x P(x)$ (Discrete RV)

(2) $E(X) = \int_{-\infty}^{\infty} x f(x) dx$ (cont. RV)

(3) $-\infty < E(x) < \infty$

(4) $E(g(x)) = \sum g(x) P(x)$ (discrete)
 $= \int g(x) f(x) dx$ (cont.)

(5) $E(C) = C$

* $E(X) = E_1 P(A) + E_2 P(A^c)$

(6) $E(aX) = aE(X)$
 For ques like coins & tossed until HH occurs
 Line recursion

(7) $E(ax+b) = aE(X) + b$

(8) $E(X+Y) = E(X) + E(Y)$

(9) $E(X-Y) = E(X) - E(Y)$

(10) $E(aX^n) = aE(X^n)$

(11) $E(XY) = E(X) \cdot E(Y/x)$

\hookrightarrow If X & Y are independent
 $E(XY) = E(X) \cdot E(Y)$ \hookrightarrow may or not

(12) $E(X^2) \neq (E(X))^2$

(13) $E(1/x) \neq 1/E(X)$

Standard Deviation

(σ)

$\sigma = \sqrt{\text{variance}}$

σ is always +ve
 i.e. ≥ 0

Variance

- ① $V(X) = E(X^2) - [E(X)]^2 = E[(X - E(X))^2]$
- ② It's always +ve. i.e. ≥ 0 .
- ③ $V(C) = 0$
- ④ $V(aX) = a^2 V(X)$
- ⑤ $V(-X) = V(X)$
- ⑥ $V(aX + b) = a^2 V(X)$
- ⑦ $V(X + Y) = V(X) + V(Y) + 2 \text{Cov}(X, Y)$
 $V(X - Y) = V(X) + V(Y) - 2 \text{Cov}(X, Y)$
→ correlation b/w X & Y
- ⑧ If X & Y are independent RV,
 $\text{Cov}(X, Y) = 0$
 $V(X + Y) = V(X) + V(Y)$
 $V(X - Y) = V(X) + V(Y)$
- ⑨ By default consider independent

$E(e^{tx}) = 1 + tx + \frac{t^2 x^2}{2!} + \frac{t^3 x^3}{3!} + \dots$
Moment Generating fun:

$E(X) \rightarrow$ 1st moment = Mean
 $E(X^2) \rightarrow$ 2nd moment

$$E(X^n) = \left[\frac{d^n}{dt^n} E(e^{tx}) \right]_{t=0}$$

PROBABILITY DISTRIBUTION

Discrete dist.

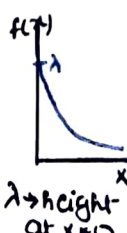

- Binomial
- Bernoulli
- Hypergeometric
- Poisson
- Geometric

Cont. dist.

- Uniform
- Exponential
- Normal

* Any probm which can be solved by Binomial can also be solved by Poisson.

Binomial Distribution	Bernoulli Distr.	Hypergeometric Distr.	Poisson Distr.	Geometric Distr.
<ul style="list-style-type: none"> Discrete RV $P + q = 1$ Independent events $P(X) = {}^n C_x p^x q^{n-x}$ $E(X) = np$ $V(X) = npq$	<ul style="list-style-type: none"> same $n = 1$ $P(X) = p^x q^{1-x}$ $E(X) = p$ $V(X) = pq$	<ul style="list-style-type: none"> Discrete RV $P + q = 1$ $P \neq q$ not independent $P = \frac{\text{Required}}{\text{Total}}$ Without replacement If we take with replacement, $p \neq q$ will become independent hence it's binomial dist. 	<small>→ $e = 2.718$</small> <ul style="list-style-type: none"> Discrete RV $P + q = 1$ Independent events $n \rightarrow \infty, p \rightarrow 0$ $P(X) = \frac{e^{-\lambda} \lambda^x}{x!}$ $x \geq 0, \lambda = np$ $E(X) = V(X) = \lambda$ (X & λ should talk about same thing) success or failure 	<ul style="list-style-type: none"> same NO combination position is fixed $P(X) = p^x q^{x-1}$ $E(X) = \frac{1}{p}$ $V(X) = \frac{q}{p^2}$

Uniform dist	Exponential dist	Normal/Gaussian dist	
<p>PDF:</p> $f(x) = \frac{1}{b-a}$ $a \leq x \leq b$ $E(X) = \frac{a+b}{2} = \frac{f(a)+f(b)}{2}$ $V(X) = \frac{(b-a)^2}{12}$	$f(x) = \lambda e^{-\lambda x}$ $x \geq 0$ $E(X) = \frac{1}{\lambda}$ $V(X) = \frac{1}{\lambda^2}$ $E(X^n) = \frac{n!}{\lambda^n}$ 	$f(x) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$ Mean = μ SD = σ <ul style="list-style-type: none"> Bell shaped curve symmetric about $x = \text{mean}$ $-\infty \int f(x) dx = \infty \int f(x) dx = \frac{1}{2}$ 	$f(x)$ is max at $x = \mu$ $f_{\max} = \frac{1}{\sqrt{2\pi}\sigma}$

Z distr (Standard Normal distr.)

To solve prob ques, normal curve is converted to Z distr
 ⑤ $f(Z) = \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}Z^2}$

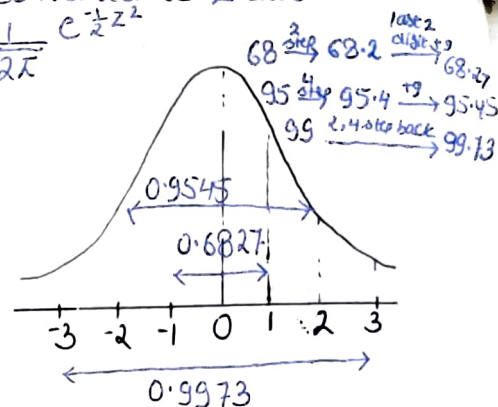
① $Z = \frac{x - \mu}{\sigma}$

② $P(Z > 0) = P(Z < 0) = \frac{1}{2}$

③ $\mu = 0, \sigma = 1$ (on comparing)

④ If x & y are independent Normal random variables then

$(ax \pm by)$ is also a normal random variable



STATISTICS

* In terms of prob, median is a point m such that $P(-\infty < X < m) = P(m < X < \infty)$

A. MEAN

① Ungrouped data

$$\text{Mean} = \frac{\sum x_i}{n}$$

② Grouped data

$$\text{Mean} = \frac{\sum x_i f_i}{\sum f_i}$$

③ If x_i given in interval find mean of class interval

B. MEDIAN

① Ungrouped data

• Arrange in Asc/Oes order

• If n is odd,

$$\text{median} = \frac{(n+1)}{2} \text{th term}$$

• If n is even,

median = mean of $\frac{n}{2}$ th, $(\frac{n}{2} + 1)$ th term

② Grouped data

use cumulative freq.

③ If x_i given in interval & N is even

$$\text{Median} = l_m + \frac{(\frac{N}{2} - \text{PCF})h}{f_m}$$

\swarrow lower limit of median class \swarrow prev. cum. freq. \swarrow freq. of median class \swarrow height of class

C. MODE

x_i having highest freq.

NOTE: {1, 2, 3, 4} No Mode

① If x_i is given in range

$$\text{Mode} = l + \frac{(f_1 - f_0)}{(2f_1 - f_0 - f_2)} \times h$$

\swarrow lower limit of modal class \swarrow freq. of modal class \swarrow height of modal class

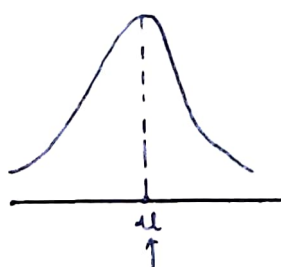
• Deane = Dorian (not always)

$$1 \text{ Mode} + 2 \text{ Mean} = 3 \text{ Median}$$

• SD, $\sigma = \sqrt{E(X^2) - (E(X))^2}$

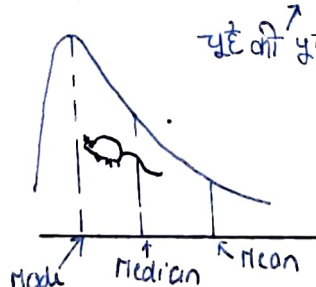
$$\sigma = \sqrt{\frac{\sum x_i^2 f_i}{\sum f_i} - \left(\frac{\sum x_i f_i}{\sum f_i}\right)^2}$$

Normal distr.



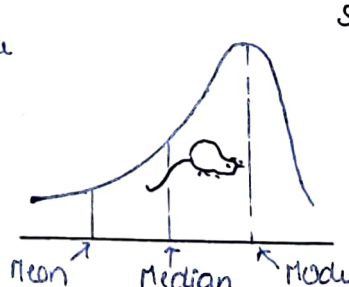
$$\text{Mean} = \text{Median} = \text{Mode}$$

Right skewed / +vely skewed



$$\text{Mean} \geq \text{Median} \geq \text{Mode}$$

Left skewed / -vely skewed



$$\text{Mode} \geq \text{Median} \geq \text{Mean}$$

• coeff of variation = $\frac{\sigma}{\mu}$ \leftarrow SD
 $\mu \leftarrow$ Mean
 • It gives amount of fluctuation in data.

NOTE: In Normal distr, if we have to find prob at a point
 $P(X=16) = P(15.5 < X < 16.5)$
 Find this area

Correlation:

r : correlation coefficient / index

① $r \in [-1, 1]$

tells nature & strength of relⁿ b/w data sets

② Direct / +ve corr.

$r = 0$ to 1

$x \uparrow \quad y \uparrow$
 $x \downarrow \quad y \downarrow$

③ Inverse / -ve corr.

$r = -1$ to 0

$x \uparrow \quad y \downarrow$
 $x \downarrow \quad y \uparrow$

Perfect -ve corr

NO corr

Perfect +ve corr

Strong +ve

Weak corr

Strong -ve

$$r = \frac{n \sum xy - \sum x \sum y}{\sqrt{n \sum x^2 - (\sum x)^2} \sqrt{n \sum y^2 - (\sum y)^2}}$$

⑤ $r = \frac{\text{COV}(X, Y)}{\sigma_x \sigma_y}$

$$\text{COV}(X, Y) = \sum (x - \bar{x})(y - \bar{y})$$

$$= E[(x - E(x))(y - E(y))] \quad n$$

x & y are two data sets

NOTE: AGP fibonacci series

$$1 \cdot x^0 + 1 \cdot x^1 + 2 \cdot x^2 + 3 \cdot x^3 + 5 \cdot x^4 + 8 \cdot x^5 + \dots = \frac{1}{1 - x - x^2}$$

Regression:

① $x = f(y)$

x on y

$y = f(x)$

y on x

② y on x

$y = a + bx$

(do \sum)

$\sum y = na + b \sum x \quad \text{--- (1)}$

$\sum xy = a \sum x + b \sum x^2 \quad \text{--- (2)}$

From eq (1) & (2) find a & b

③ Dividing eq (1) by n

$\bar{y} = a + b \bar{x}$

\bar{x} & \bar{y} also lie on regression line.

$n \rightarrow$ NO. of ele in a data set

④ y on x

y par to x based on

$(y - \bar{y}) = by_x (x - \bar{x})$

$by_x = \frac{n \sum xy - \sum x \sum y}{n \sum x^2 - (\sum x)^2}$

(last mb \rightarrow x so deno me sb x)

⑤ by_x is same as b in Method 1

⑥ x on y

x par to y based on

$(x - \bar{x}) = bx_y (y - \bar{y})$

$bx_y = \frac{n \sum xy - \sum x \sum y}{n \sum y^2 - (\sum y)^2}$

(last mb \rightarrow y so deno me sb y)

⑦ Relⁿ b/w correlation & Regression coeff.

(i) $bx_y = r \frac{\sigma_x}{\sigma_y}$

$by_x = r \frac{\sigma_y}{\sigma_x}$

sign same

(ii) $bx_y * by_x = r^2$

\Rightarrow convolution theorem

Let x, y be random var

$\text{var} \leftarrow P_x, P_y$ be PDF.

$Z = x + y$

$P_z(z) = \sum_x P_x(x) P_y(z - x)$

(for discrete RV)

$f_z(z) = \int_{-\infty}^{\infty} f_x(x) f_y(z - x) dx$

(for continuous RV)

* If x_1 and x_2 are independent exponential random var. with parameters λ_1 & λ_2 then $x = \min(x_1, x_2)$ is expo. RV with parameter $\lambda_1 + \lambda_2$.

* A ball is picked from bag & placed back along with 1 more ball of same colour every trial. Then, $P(\text{red ball})$ is same in every iteration. (GATE 21)