

Sigma algebra

A collection of subsets of S is called a sigma algebra, denoted by \mathcal{B} , if it satisfies the following properties

$$(i) \emptyset \in \mathcal{B}$$

$$(ii) \text{ if } A \in \mathcal{B} \text{ Then } A^c \in \mathcal{B}$$

$$(iii) \text{ if } A_1, A_2, \dots \in \mathcal{B} \text{ Then } \bigcup_{i=1}^{\infty} A_i \in \mathcal{B}$$

(closed under countable unions)

e.g.

$$\mathcal{B} = \{ \emptyset, S \} \quad \text{--- trivial sigma algebra}$$

if S is finite Then $\mathcal{B} = \{ \text{all subsets of } S, \text{ including } S \text{ itself} \}$

Probability function

Given a sample space S and an associated sigma algebra \mathcal{B} , a probability function P with domain

\mathcal{B} , That satisfies

is a Prob. function

$$(i) P(A) \geq 0 \text{ for all } A \in \mathcal{B}$$

$$(ii) P(S) = 1$$

$$(iii) \text{ if } A_1, A_2, \dots \in \mathcal{B} \text{ are pairwise disjoint}$$

$$\text{Then } P\left(\bigcup_{i=1}^{\infty} A_i\right) = \sum_{i=1}^{\infty} P(A_i)$$

Kolmogorov's
Axioms

Axioms of
Probability

Example. Consider a simple experiment of tossing a fair coin. (30)

Coin.

$$\text{So } S = \{H, T\}$$

$$P(\{H\}) = P(\{T\}) \quad \rightarrow (1) \quad \left. \begin{array}{l} \text{given fair} \\ \text{coin} \end{array} \right\}$$

$$\text{Since } S = \{H\} \cup \{T\}$$

$$P(S) = P(\{H\} \cup \{T\})$$

$$1 = P(\{H\}) + P(\{T\}) \quad \rightarrow (2)$$

$$\text{Solving (1) \& (2) gives } P(\{H\}) = P(\{T\}) = \frac{1}{2}$$

if not give fair coin, any function P

that satisfies (2) can be considered

as Probability function

Th. if P is a probability function and A is any set in \mathcal{B} Then

$$(i) P(\emptyset) = 0 \quad (ii) P(A) \leq 1 \quad (iii) P(A^c) = 1 - P(A)$$

Proof

Since $A \cup A^c$ forms a partition of the sample space

$$S = A \cup A^c \Rightarrow P(S) = P(A \cup A^c)$$

$$1 = P(A) + P(A^c) \Rightarrow P(A^c) = 1 - P(A) \quad \left. \begin{array}{l} \text{of (iii) axioms} \\ \vdots \end{array} \right\}$$

$$\text{Since } A \in \mathcal{B} \Rightarrow A^c \in \mathcal{B}$$

$$P(A^c) \geq 0 \Rightarrow 1 - P(A) \geq 0 \Rightarrow P(A) \leq 1 \quad \left. \begin{array}{l} \text{of (ii) axioms} \\ \vdots \end{array} \right\}$$

$$\text{Also } S = S \cup \emptyset \Rightarrow P(S) = P(S \cup \emptyset)$$

$$1 = P(S) + P(\emptyset) \Rightarrow P(\emptyset) = 0$$

Th. if P is a probability function, and A and B are any sets in \mathcal{B} Then

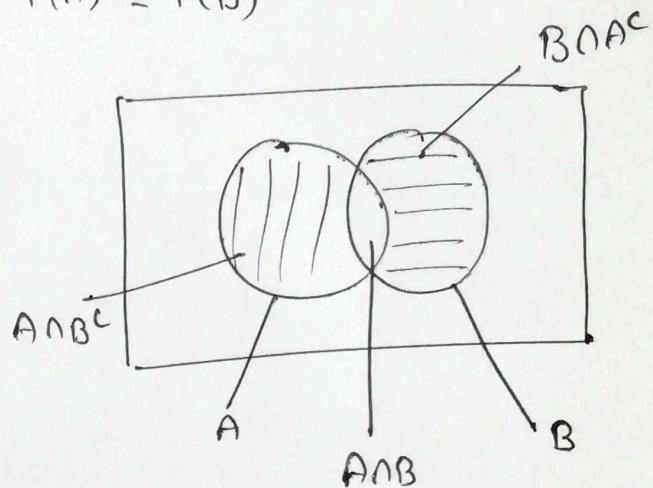
$$(i) P(B \cap A^c) = P(B) - P(A \cap B)$$

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

$$(iii) \text{ if } A \subset B \text{ Then } P(A) \leq P(B)$$

Proof.

$$(i) B = \{B \cap A\} \cup \{B \cap A^c\}$$



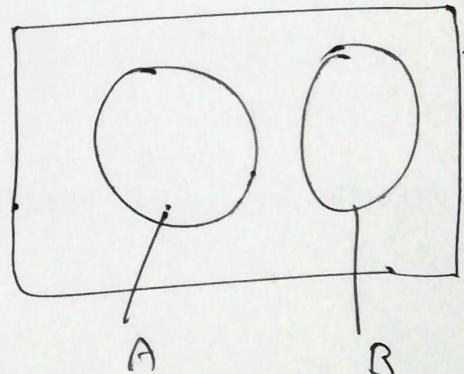
$$P(B) = P(\{B \cap A\} \cup \{B \cap A^c\})$$

$$= P(B \cap A) + P(B \cap A^c)$$

$\therefore \{B \cap A\}$ and $\{B \cap A^c\}$ are disjoint

$$\left. \begin{aligned} & \therefore P(B \cap A^c) = P(B) - P(A \cap B) \\ & \end{aligned} \right\} \quad (1)$$

$$\text{Also } (ii) A \cup B = A \cup \{B \cap A^c\}$$



$$P(A \cup B) = P(A \cup \{B \cap A^c\})$$

$$= P(A) + P(B \cap A^c)$$

$$\left. \begin{aligned} & = P(A) + P(B) - P(A \cap B) \\ & \end{aligned} \right\} \quad \text{by (1)}$$

$$(iii) \text{ if } A \subset B \text{ Then } A \cap B = A$$

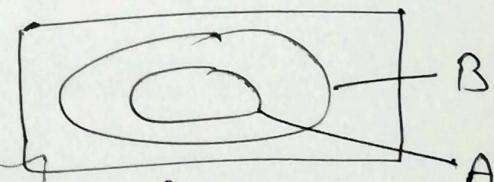
$$\therefore \text{ by (1)} \quad P(B \cap A^c) = P(B) - P(A \cap B)$$

$$= P(B) - P(A)$$

then know $P(B \cap A^c) \geq 0$

$$P(B) - P(A) \geq 0$$

$$P(B) \geq P(A) \Rightarrow \left\{ P(A) \leq P(B) \right\}$$



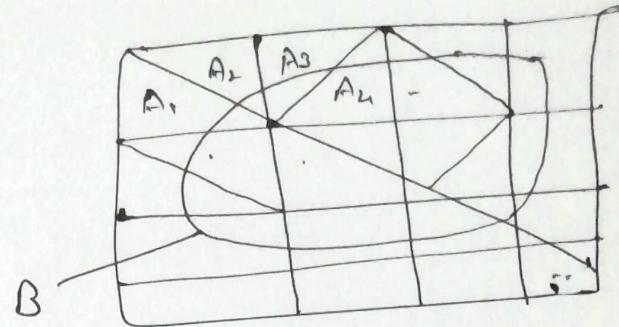
Th⁴: if P is a probability function, Then

$$P(B) = \sum_{i=1}^{\infty} P(B \cap A_i) \text{ for any partition } A_1, A_2, \dots$$

Proof

$$B = B \cap S$$

$$B = B \cap (\bigcup_{i=1}^{\infty} A_i)$$



Since A_i is any partition

$$\therefore S = \bigcup_{i=1}^{\infty} A_i \quad \text{and} \quad A_i \cap A_j = \emptyset \text{ for all } i \neq j$$

Therefore

$$P(B) = P(B \cap (\bigcup_{i=1}^{\infty} A_i))$$

$$= P\left(\bigcup_{i=1}^{\infty} (B \cap A_i)\right)$$

$$P(B) = \sum P(B \cap A_i)$$

\therefore of (iii) axiom

} Since A_i 's are disjoint, The sets $B \cap A_i$ are also disjoint

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Conditional Probability if A and B are events in S

and $P(B) > 0$ Then The Conditional Probability of A given B , written $P(A|B)$, is

$$P(A|B) = \frac{P(A \cap B)}{P(B)}$$

$$P(A \cap B) = P(B) P(A|B)$$

$$P(A \cap B) = P(A) P(B|A)$$

Baye's Rule

Let A_1, A_2, \dots be a partition of the sample space and let B be any set. Then, for each $i=1, 2, \dots$

$$P(A_i | B) = \frac{P(A_i) P(B|A_i)}{\sum_{i=1}^{\infty} P(A_i) P(B|A_i)}$$

Proof

$$\text{Take L.H.S } P(A_i | B) = \frac{P(A_i \cap B)}{P(B)}$$

$$= \frac{P(A_i) P(B|A_i)}{\sum_{j=1}^{\infty} P(B \cap A_j)}$$

$$= \frac{P(A_i) P(B|A_i)}{\sum_{j=1}^{\infty} P(A_j) P(B|A_j)}$$

= R.H.S

↑
Law of
total Probability

Example A factory uses three machines X, Y, Z to produce certain items. Suppose

- (i) Machine X produces 50 percent of the items of which 3 percent are defective
- (ii) Machine Y produces 30 percent of the items of which 4 percent are defective
- (iii) Machine Z produces 20 percent of the items of which 5 percent are defective.

(34)

Then find

1. The probability p that a randomly selected item is defective
2. Suppose a defective item is found among the output
find the probability it come from

2.1 machine X

2.2 machine Y

2.3 machine Z

Sol.

(i) Let D : denote the event that an item is defective

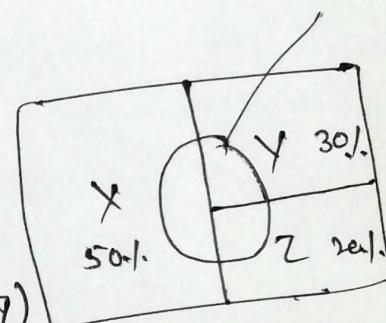
$$D = \{D \cap X\} \cup \{D \cap Y\} \cup \{D \cap Z\}$$

$$P(D) = P(D \cap X) + P(D \cap Y) + P(D \cap Z)$$

$$= P(X)P(D|X) + P(Y)P(D|Y) + P(Z)P(D|Z)$$

$$= 0.50 \times 0.03 + 0.30 \times 0.04 + 0.20 \times 0.05$$

$$= 0.037 = 3.7\%$$



Partition total — 100%

$$(2.1) \quad P(X|D) = \frac{P(X \cap D)}{P(D)} = \frac{P(X)P(D|X)}{P(D)} = \frac{0.50 \times 0.03}{0.037} = 40.5\%$$

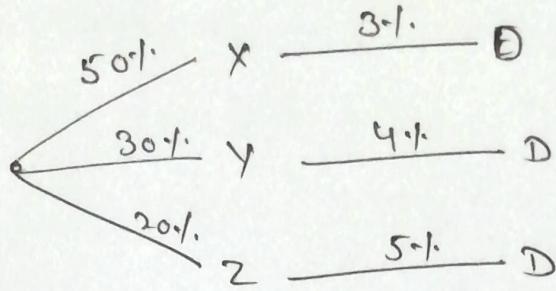
$$(2.2) \quad P(Y|D) = \frac{P(Y \cap D)}{P(D)} = \frac{P(Y)P(D|Y)}{P(D)} = \frac{0.30 \times 0.04}{0.037} = 32.5\%$$

$$(2.3) \quad P(Z|D) = \frac{P(Z \cap D)}{P(D)} = \frac{P(Z)P(D|Z)}{P(D)} = \frac{0.20 \times 0.05}{0.037} = 27.5\%$$

Alternatively

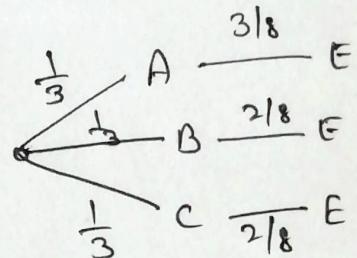
Using tree diagram

$$\begin{aligned} P(D) &= 0.50 \times 0.03 \\ &\quad + 0.30 \times 0.04 \\ &\quad + 0.20 \times 0.05 \\ &= 0.037 \\ &= 3.7\% \end{aligned}$$



$$P(X|D) = \frac{0.65 \times 0.03}{0.037} = 40.5\%.$$

$\underline{\underline{}}$



Q3

Suppose the following three boxes are given:

Box A contains 3 Red and 5 White Balls

Box B contains 2 Red and 1 White ball

Box C contains 2 Red and 3 White balls

A Box is selected at random, and a ball is drawn from the box. If the ball is red, find the probability that it came from box A.

Soln.

$$P(A) = P(B) = P(C) = \frac{1}{3}$$

Selecting a box A, B, C

E: Selecting a red ball

$$P(E|A) = \frac{3c_1}{8c_1} = \frac{3}{8}$$

$$P(E|B) = \frac{2c_1}{8c_1} = \frac{2}{8}$$

$$P(E|C) = \frac{2c_1}{8c_1} = \frac{2}{8}$$

$$\begin{aligned} P(A|E) &= \frac{P(A \cap E)}{P(E)} \\ &= \frac{P(A)P(E|A)}{P(E)} \\ &= \frac{\frac{1}{3} \times \frac{3}{8}}{\frac{173}{360}} \\ &\approx 0.26 \end{aligned}$$

$$\begin{aligned} P(E) &= P(A)P(E|A) + P(B)P(E|B) + P(C)P(E|C) \\ &= \frac{1}{3} \times \frac{3}{8} + \frac{1}{3} \times \frac{2}{8} + \frac{1}{3} \times \frac{2}{8} = \frac{173}{360} \approx 0.48 \end{aligned}$$

Statistically Independent

Two events, A and B, are statistically independent if $P(A \cap B) = P(A)P(B)$

Th. if A and B are independent events, Then
The following pairs are also independent:

$$(i) A \text{ and } B^c$$

$$(ii) A^c \text{ and } B$$

$$(iii) A^c \text{ and } B^c$$

Proof

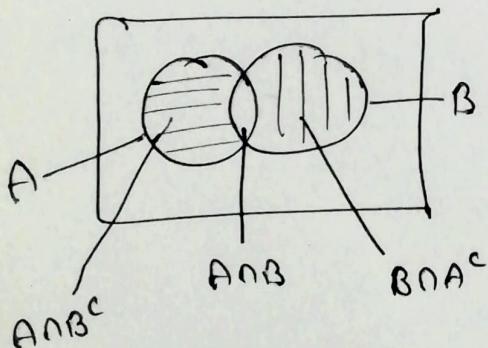
Given A and B are independent events

$$\therefore P(A \cap B) = P(A)P(B)$$

To show (i) A and B^c are independent

$$\text{we have to show } P(A \cap B^c) = P(A)P(B^c)$$

$$\text{Now, we know } P(A \cap B^c) = P(A) - P(A \cap B)$$



$$= P(A) - P(A)P(B)$$

$$= P(A) [1 - P(B)]$$

$$= P(A) P(B^c)$$

$$(ii) P(A^c \cap B) = P(B) - P(A \cap B) = P(B) - P(A)P(B) = P(A^c)P(B)$$

$$(iii) P(A^c \cap B^c) = P((A \cup B)^c) = 1 - P(A \cup B)$$

$$= 1 - P(A) - P(B) + P(A \cap B)$$

$$= P(A^c) - P(B) + P(A)P(B) = P(A^c) - P(B)[1 - P(A)]$$

$$= P(A^c) [1 - P(B)] = P(A^c)P(B^c)$$

We might think that we could say A, B and C are independent

$$\text{if } P(A \cap B \cap C) = P(A)P(B)P(C)$$

However, This is not the correct condition.

Eg

Let an experiment, consist of tossing two dice

Define following events

$$A = \{\text{double appears}\}$$

$$B = \{\text{The sum is between 7 and 10}\}$$

$$C = \{\text{The sum is 2 or 7 or 8}\}$$

$$\text{Now } P(A) = \frac{1}{6}$$

$$P(B) = \frac{1}{2}$$

$$P(C) = \frac{1}{3}$$

$$\text{Also } P(A \cap B \cap C)$$

$$= P(\text{Sum is 8, Composed of double 4's})$$

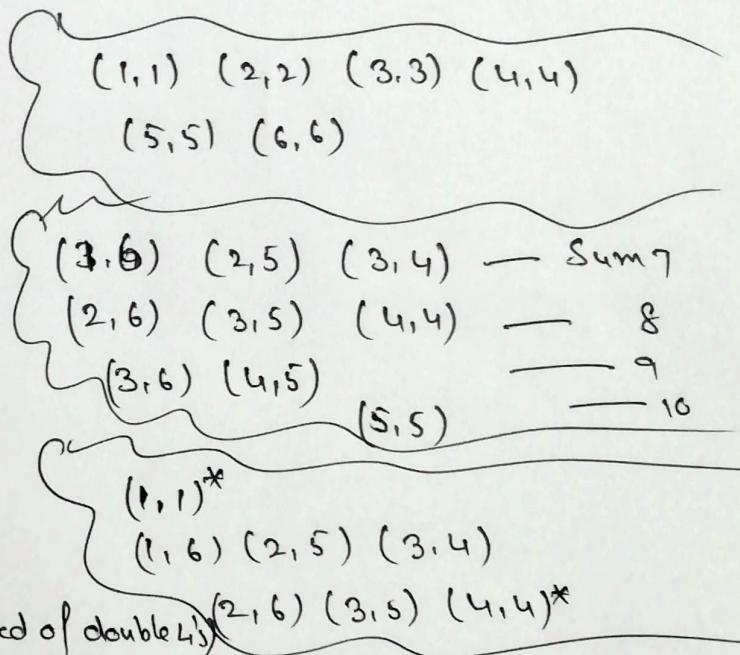
$$= \frac{1}{36}$$

$$= \frac{1}{6} \times \frac{1}{2} \times \frac{1}{3} = P(A)P(B)P(C)$$

$$\text{However, } P(B \cap C) = P(\text{sum equals 7 or 8})$$

$$= \frac{11}{36} \neq P(B)P(C)$$

$$\text{Similarly } P(A \cap B) \neq P(A)P(B)$$



Not a strong
condition to
guarantee
Pairwise independence

Definition: A collection of events A_1, A_2, \dots, A_n

are mutually independent if for any

Subcollection A_{i_1}, \dots, A_{i_k} , we have

$$P\left(\bigcap_{j=1}^k A_{i_j}\right) = \prod_{j=1}^k P(A_{i_j})$$

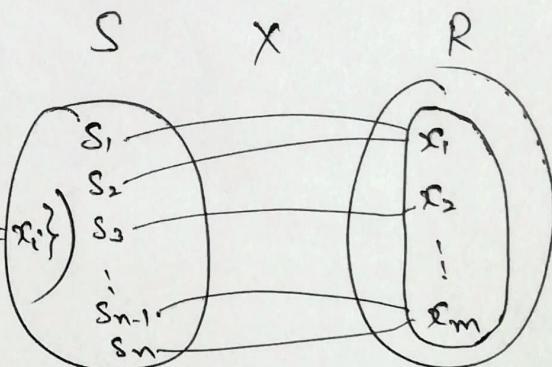
Random Variable

A random variable is a function from a sample space S into the real numbers.

} easy to deal with a summary variable than with the original probability structure

$$x: S \rightarrow R$$

$$P_x(x = x_i) = P(\{s_j \in S : x(s_j) = x_i\})$$



or we can write

$$P(x = x_i) = P(\{s_j \in S : x(s_j) = x_i\})$$

$$P(x = x_1) = P(\{s_1, s_2\})$$

$$P(x = x_2) = P(\{s_3\})$$

$$P(x = x_m) = P(\{s_{m-1}, s_m\})$$

observe
carefully

Example

Consider an experiment of tossing a fair coin Three times

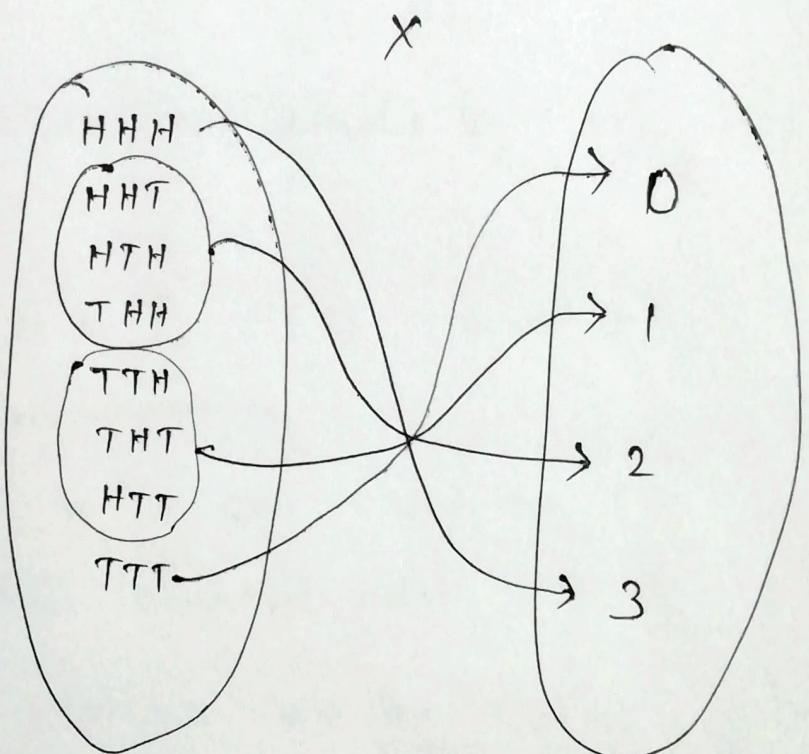
Define The random variable X to be The number of heads obtained in Three tosses.

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

X : No. of heads obtained in Three tosses

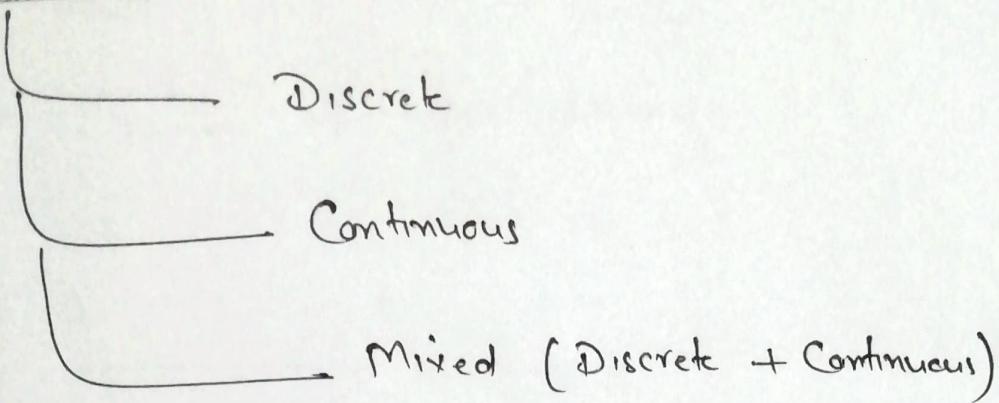
Range of r.v. X

$$\text{is } X = \{0, 1, 2, 3\}$$



S	HHH	HHT	HTH	THH	TTH	THT	HTT	TTT
$X(s)$	3	2	2	2	1	1	1	0

x	0	1	2	3
$P(X=x)$	$\frac{1}{8}$	$\frac{3}{8}$	$\frac{3}{8}$	$\frac{1}{8}$

Random VariableCumulative Distribution function (CDF)

The CDF of a r.v., x , denoted by $F_x(x)$,
is defined by

$$\underline{F_x(x) = P_x(x \leq x)}, \text{ for all } x$$

To: The function $F(x)$ is a CDF if and only if
The following Three conditions hold:

(i) $\lim_{x \rightarrow -\infty} F(x) = 0$ and $\lim_{x \rightarrow \infty} F(x) = 1$

(ii) $F(x)$ is non-decreasing function of x

(iii) $F(x)$ is right-continuous; That is,

for every number x_0 , $\lim_{x \rightarrow x_0} F(x) = F(x_0)$

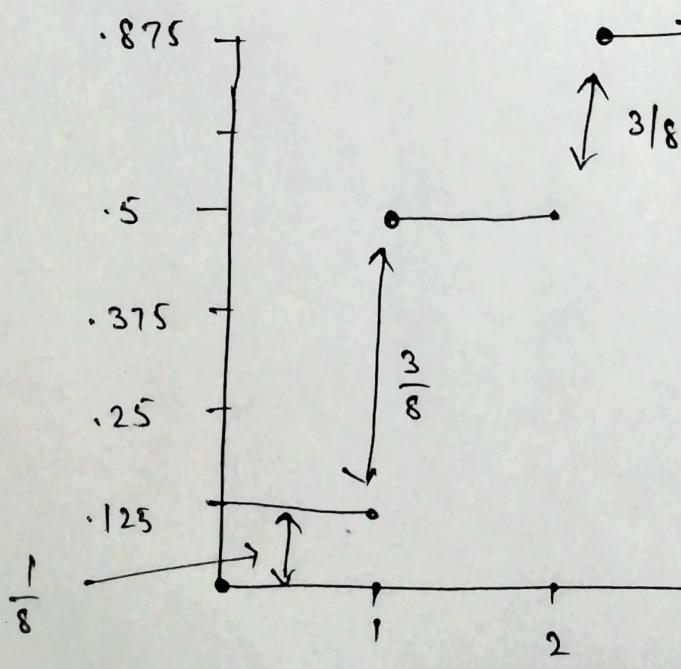
Eg Consider The example of tossing a fair coin
Three times

(41)

X : number of heads observed

X	0	1	2	3
$P(X=x)$	$\frac{1}{8}$	$\frac{3}{8}$	$\frac{3}{8}$	$\frac{1}{8}$
$F(x)$	$\frac{1}{8}$	$\frac{4}{8}$	$\frac{7}{8}$	1

$$F(x) = \begin{cases} 0 & \text{if } -\infty < x < 0 \\ \frac{1}{8} & \text{if } 0 \leq x < 1 \\ \frac{1}{2} & \text{if } 1 \leq x < 2 \\ \frac{7}{8} & \text{if } 2 \leq x < 3 \\ 1 & \text{if } 3 \leq x < \infty \end{cases}$$



Graph is

Step-function

Size of The jump at $x_i \in X$ is equal to $P(X=x_i)$