

Random Variable

A random variable (x) is defined as a function that associates a real number (the probability value) to an outcome of an experiment. $X: S \rightarrow N \{0, 1, 2, 3\}$

Ex- Tossing two fair coin,

$$\text{sample space} = \left\{ \underset{2}{HH}, \underset{1}{HT}, \underset{1}{TH}, \underset{0}{TT} \right\}$$

$$X = (\text{count the number of time Head turns}) \\ = 2, 1, 0$$

What is the probability that X can take the value 2,
 $P(X=2) = P(HH) = 1/4$

What " " " 1,
 $P(X=1) = P(TH+HT) \Rightarrow 1/4 + 1/4 \Rightarrow 1/2$

What " " " 0,
 $P(X=0) = P(TT) \Rightarrow 1/4$

Ex- Flipping three fair coin;

$$SS = \left\{ \underset{1/8}{HHH}, \underset{1/8}{HHT}, \underset{1/8}{HTH}, \underset{1/8}{HTT}, \underset{1/8}{TTH}, \underset{1/8}{THT}, \underset{1/8}{TTH}, \underset{1/8}{TTT} \right\}$$

$$X = (\text{defn no of Tail come out})$$

$$(X) = 3, 2, 1, 0$$

→ random variable.

What is the probability that X can take the value 0;
 $P(X=0) = P(HHH) = 1/8$

$$P(X=1) = P(HHT + HTH + TTH) \\ = 3/8$$

} probability distribution for random variable.

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

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

Ex-3 Rolling two dice.

$$\text{Sample space} = \left\{ \underset{\frac{1}{36}}{(1,1)} \underset{\frac{1}{36}}{(1,2)} \dots \dots \dots \underset{\frac{1}{36}}{(6,6)} \right\}$$

$$X = (\text{sum values on two dice})$$

$$= \{2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12\}$$

→ Random value.

$$P(X=2) = \frac{1}{36}$$

$$P(X=3) = P\{(1,2) (2,1)\} = \frac{2}{36}$$

$$P(X=4) = P\{(1,3) (2,2) (3,1)\} = \frac{3}{36}$$

$$P(X=12) = P\{(6,6)\} = \frac{1}{36}$$

Another example for the Random variable:

→ Example 1: **Random variable**

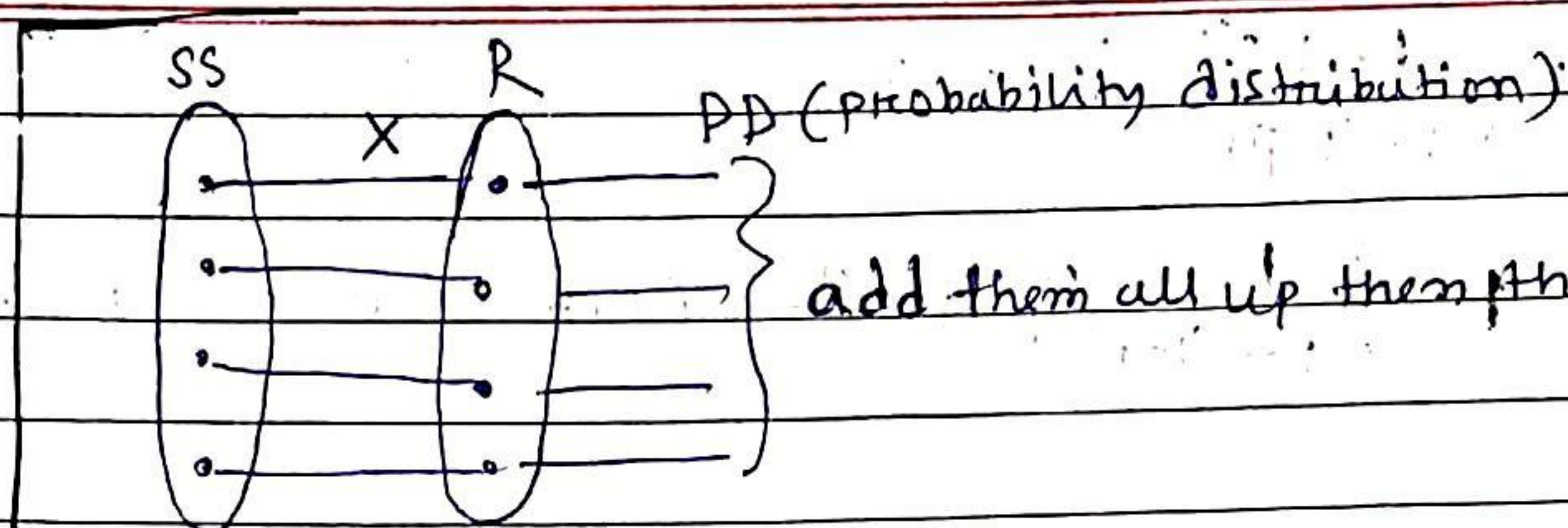
(Example 2)

Discrete

Random variable

Continuous

Random variable



- ~~Example~~ A fair coin flip twice - $SS = \{HH, HT, TH, TT\}$
 $P(X=0) = \frac{1}{4}$ $X = (\text{define no of time head come out})$
 $P(X=1) = \frac{2}{4}$ $(\frac{1}{4} + \frac{2}{4} + \frac{1}{4} \Rightarrow \frac{4}{4} \Rightarrow 1)$
 $P(X=2) = \frac{1}{4}$

Ex-4 Have a coin, then probability of getting head ' p '. and probability of getting Tail is ' $(1-p)$ '. Experiment is flipping a coin until I get a head and ' N ' denotes a random variable which indicates ^{the} number of time I have flip the coin until I get the first head.

→ $N = (\text{define no of time flipped the coin until I get the first head})$
 $N = (1, 2, 3, 4, \dots)$

Probability of getting head in first attempt -
 $P(N=1) = P(H) = p$

$$P(N=2) = P(TH) = (1-p)p$$

$$P(N=3) = P(TTH) = (1-p)(1-p)p$$

$$P(N=4) = P(TTTH) = (1-p)(1-p)(1-p)p$$

⋮

$$P(N=n) = P(TT \dots H) = (1-p)^{n-1} p$$

$$\begin{aligned} \sum_{n=1}^{\infty} P(N=n) &= p + (1-p)p + (1-p)^2 p + \dots + (1-p)^{n-1} p \\ &= p \left(\frac{1}{1-(1-p)} + (1-p) + (1-p)^2 + \dots + (1-p)^{n-1} \right) \end{aligned}$$

$$1 - (1-p)^n \approx 1$$

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

$$\sum_{n=1}^{\infty} p(H=n) \Rightarrow p \left(\frac{1}{1 - (1-p)} \right) = 1 \quad (\text{sum of all are '1'})$$

• Bernoulli random variable :-

→ In an experiment whose outcome can be classified as either a success or a failure, let 'x=1' if the outcome is success and 'x=0' if it is a failure. The probability mass function of 'x' is given by -

$$p(0) = p(x=0) = (1-p)$$

$$p(1) = p(x=1) = p, \text{ where 'p' is probability of success.}$$

x is Bernoulli random variable.

example -

1) Rolling a dice - picking up '1' - if '1' comes up then success
 $SS = \{1, 2, 3, 4, 5, 6\}$ If any other comes up then failure
 $\frac{1}{6} \quad \frac{5}{6}$

↓ success ↓ failure

$$p(x=1) = \frac{1}{6} \quad p(x=0) = \frac{5}{6}$$

In this experiment probability of getting success is $\frac{1}{6}$ and failure is $\frac{5}{6}$.

• Binomial random Variable:

→ suppose 'n' independent experiments, each of which results in a 'success' with probability p and in a failure with probability $(1-p)$ are to be performed. If 'x' represents the number of successes that occur in 'n' trials, then x is called binomial random variable with parameters (n, p) .

The probability mass function of a binomial random variable having parameters (n, p) is given by

$$P(i) = {}^n C_i p^i (1-p)^{n-i}, \quad i = 0, \dots, n$$

Ex-1 Flipping a coin, $P(H) = 1/2$ and $P(T) = 1/2$ here getting a ~~success~~ ^{Head} is success and getting a tail is failure. What is the prob that you get exactly one success.

→

$$HTTT + THTT + TTHT + TTTT$$

probability of getting 1 success is =

$$P(1) = {}^4 C_1 p (1-p)^{4-1}$$

$$= {}^4 C_1 p (1-p)^3$$

Ex-2

It is known that any item produced by a certain machine will be defective with probability 0.1, independently of any other item. What is the probability that in a sample of 3 items,

(a) exactly one is defective.

$$P(i) = {}^n C_i p^i (1-p)^{n-i}$$

$$P(x=1) \Rightarrow P(1) = {}^3 C_1 (0.1) (1-0.1)^{3-1} \Rightarrow {}^3 C_1 (0.1) (0.9)^2$$

(b) At most one

$$P(X=0) = {}^3C_0 (0.1)^0 (0.9)^3$$

$$\Rightarrow P(X=0) + P(X=1) = {}^3C_1 (0.1) (0.9)^2 + {}^3C_0 (0.1)^0 (0.9)^3$$

(c) At least one ($P(X=1) + P(X=2) + P(X=3)$)

$$P(X=1) + P(X=2) + P(X=3)$$

① Poisson random variable -

→ A random variable 'X', taking on one of the values 0, 1, 2, ... is said to be poisson random variable with parameter λ , if for some $\lambda > 0$,

$$P(i) = P(X=i) = e^{-\lambda} \frac{\lambda^i}{i!}, \quad i=0, 1, 2, 3, \dots$$

Note: $\lambda \rightarrow (n \uparrow, p \downarrow) \rightarrow (1n \text{ Binomial rev})$

$$\lambda = np$$

Binomial random variable can be approximated using poisson random variable.

Ex-1 Suppose that the number of typographical errors on a single page of this book has a poisson distribution with $\lambda=1$. Calculate the probability that there is at least one error on this page.

$$\begin{aligned} \rightarrow P(X=2) &= e^{-\lambda} \frac{\lambda^i}{i!} \\ &= e^{-1} \frac{1^2}{2!} \\ \text{prob of no error,} \\ P(0) &= e^{-1} \frac{\lambda^0}{0!} \\ &= \frac{1}{e} \end{aligned} \quad \left| \quad \begin{aligned} \rightarrow P(X=0) + P(X=1) + P(X=2) + \dots &= 1 \\ P(X \geq 1) &= 1 - P(X=0) \\ &= 1 - e^{-1} \frac{\lambda^0}{0!} \\ &= (1 - \frac{1}{e}) \end{aligned}$$

Ex-2 If the number of accidents occurring on a highway is a poisson random variable with parameter $\lambda = 3$, what is the probability that no accidents occur today.

$$\rightarrow P(i) = P(X=i) = e^{-\lambda} \frac{\lambda^i}{i!}$$

✓ for no accident - $P(X=0) = e^{-3} \frac{3^0}{0!} \Rightarrow \frac{1}{e^3}$

for one accident today, $P(X=1) = e^{-3} \frac{3^1}{1!}$
 $= \frac{3}{e^3}$

Numbers of accident today at least '1' ~~are~~

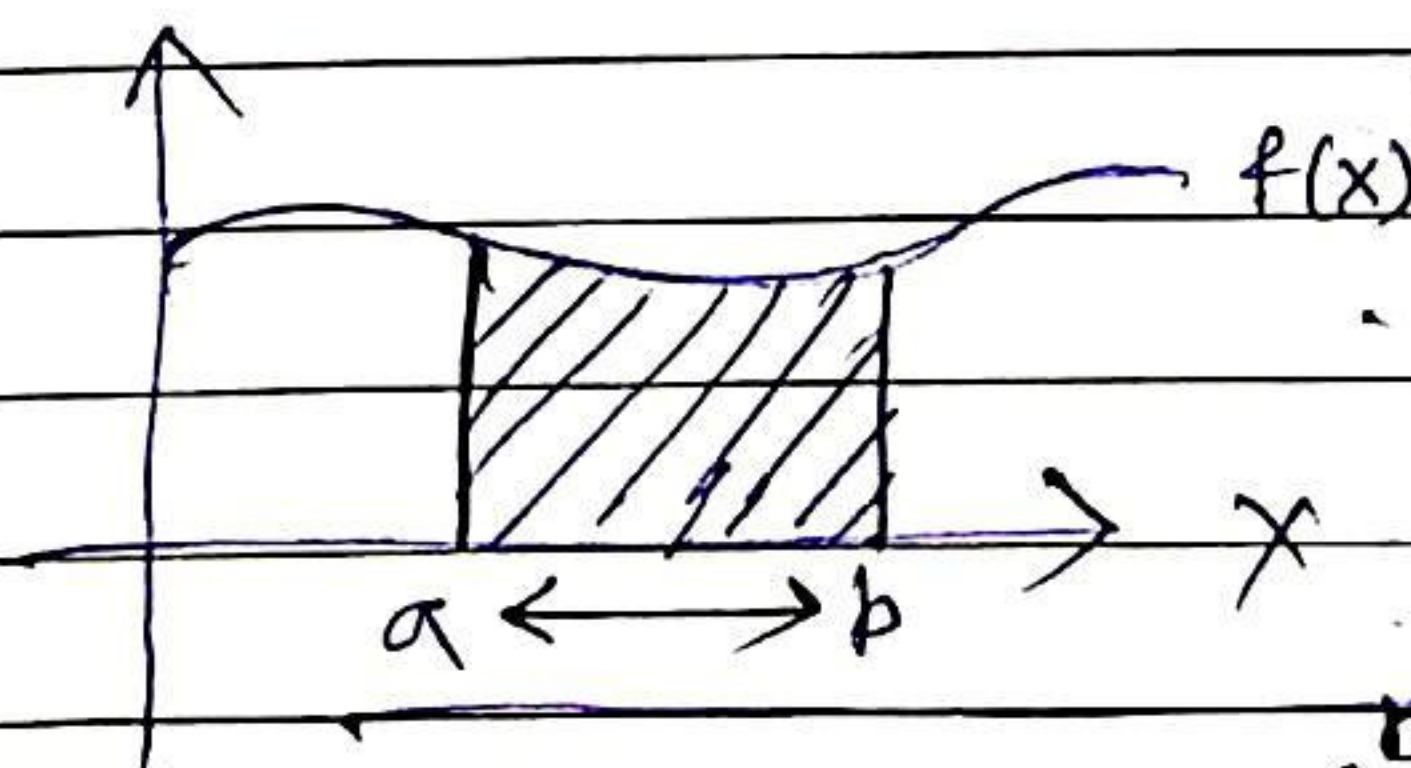
$$P(X \geq 1) = 1 - P(X=0)$$

$$= 1 - \frac{1}{e^3}$$

• Continuous Random variable → (continuous)

→ 'X' is a continuous random variable if the set of possible values it could take is uncountable, and there exists a non negative function $f(x)$, called probability density function, having ~~at~~ the property that for any set 'B' of real numbers -

$$P\{X \in B\} = \int_B f(x) \cdot dx \quad \text{1890-1920 (take a range)}$$



$$P(a \leq X \leq b) = \int_a^b f(x) dx$$

$$P(X=a) = \int_a^a f(x) dx = 0$$

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

→ Total probability under the area curve always 1.

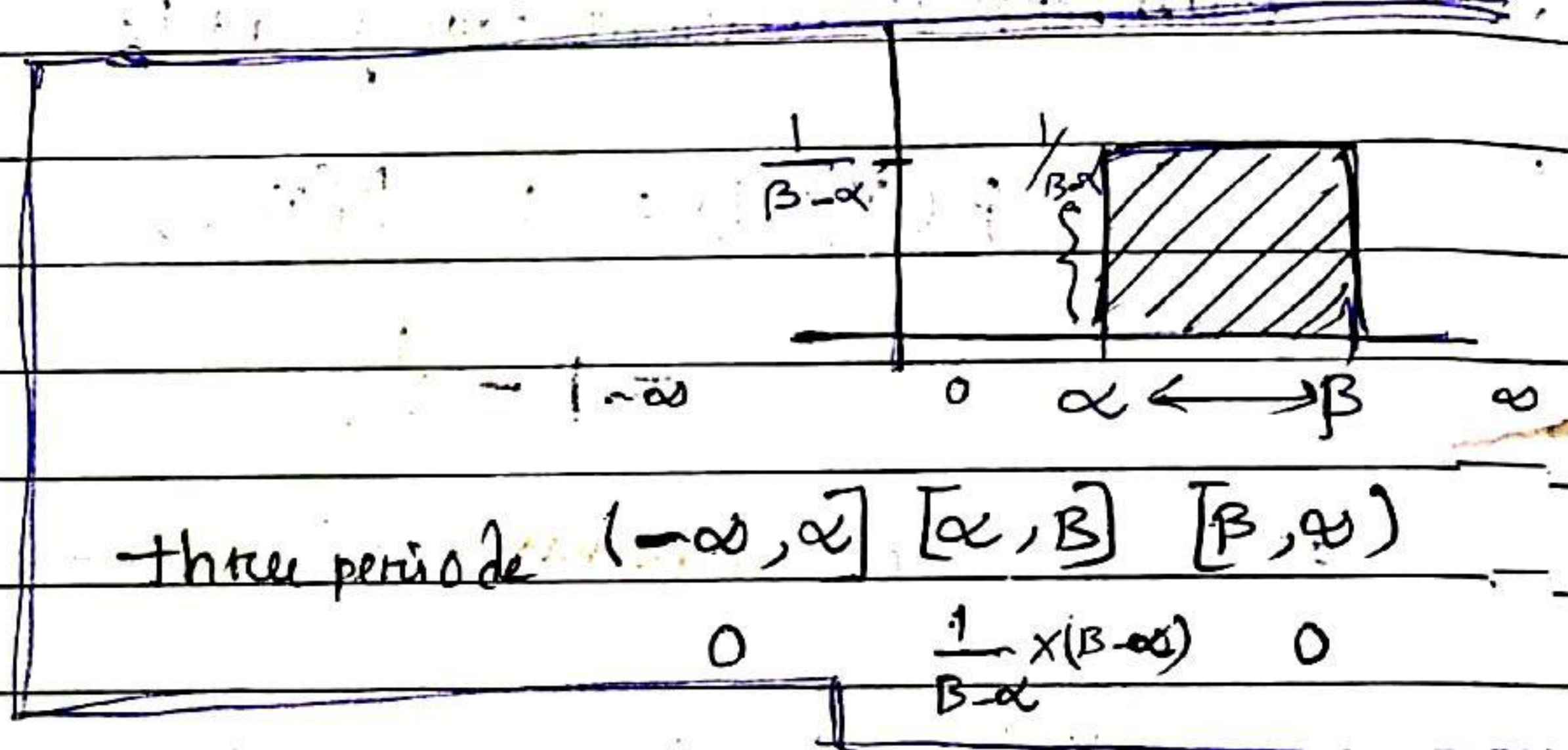
$$\int_{-\infty}^{\infty} f(x) dx = 1$$

• Uniform Random variable — (continuous) Rv

→ 'x' is a uniform random variable on the interval (α, β) if its probability density function is given

by,

$$f(x) = \begin{cases} \frac{1}{\beta - \alpha} & \text{if } \alpha < x < \beta \\ 0 & \text{otherwise.} \end{cases}$$



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

$$\Rightarrow \int_{-\infty}^{\alpha} f(x) dx + \int_{\alpha}^{\beta} f(x) dx + \int_{\beta}^{\infty} f(x) dx$$

$$\Rightarrow 0 + \int_{\alpha}^{\beta} \frac{1}{\beta - \alpha} dx + 0$$

$$\Rightarrow \frac{1}{(\beta - \alpha)} [x]_{\alpha}^{\beta}$$

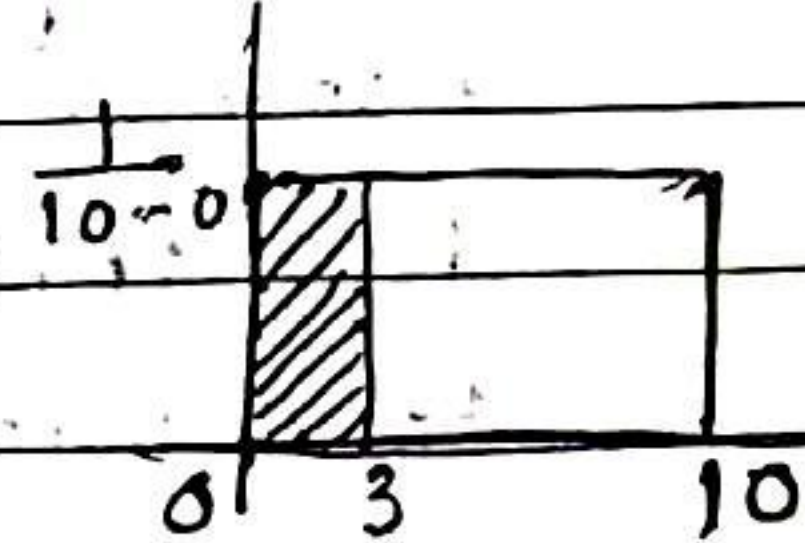
$$\Rightarrow \frac{1}{(\beta - \alpha)} (\beta - \alpha)$$

$$\Rightarrow 1$$

Ex-1 If 'x' is uniformly distributed over (0,10), calculate the probability that -

- (a) $x < 3$ (b) $x > 7$ (c) $1 < x < 6$

$$\rightarrow f(x) = \begin{cases} \frac{1}{10-0} & 0 < x < 10 \\ 0 & \text{otherwise} \end{cases}$$



$$\begin{aligned} \text{(a)} \int_{-\infty}^3 f(x) dx &\Rightarrow \int_{-\infty}^3 \frac{1}{10} dx \Rightarrow \frac{1}{10} [x]_{-\infty}^3 \\ &\Rightarrow \frac{1}{10} (3 - (-\infty)) \\ &\Rightarrow \frac{3}{10} \end{aligned}$$

$$\text{(b)} \int_7^{\infty} f(x) dx \Rightarrow \int_7^{10} \frac{1}{10} dx \Rightarrow \frac{1}{10} (10 - 7) \Rightarrow \frac{3}{10}$$

$$\text{(c)} \int_1^6 f(x) dx = \left(\frac{1}{10}\right) \times 5 \Rightarrow \frac{5}{10} = \frac{1}{2}$$

• Exponential Random variable -

→ A continuous random variable whose probability density function is given by, for some $\lambda > 0$.

$$f(x) = \begin{cases} \lambda e^{-\lambda x}, & \text{if } x \geq 0 \\ 0, & \text{if } x < 0 \end{cases}$$

$$\int_{-\infty}^{\infty} \lambda e^{-\lambda x} dx = 1$$

$$\begin{aligned} \text{Ex-} P(a \leq x \leq b) \\ = \int_a^b \lambda e^{-\lambda x} dx \end{aligned}$$

• Normal random variable -

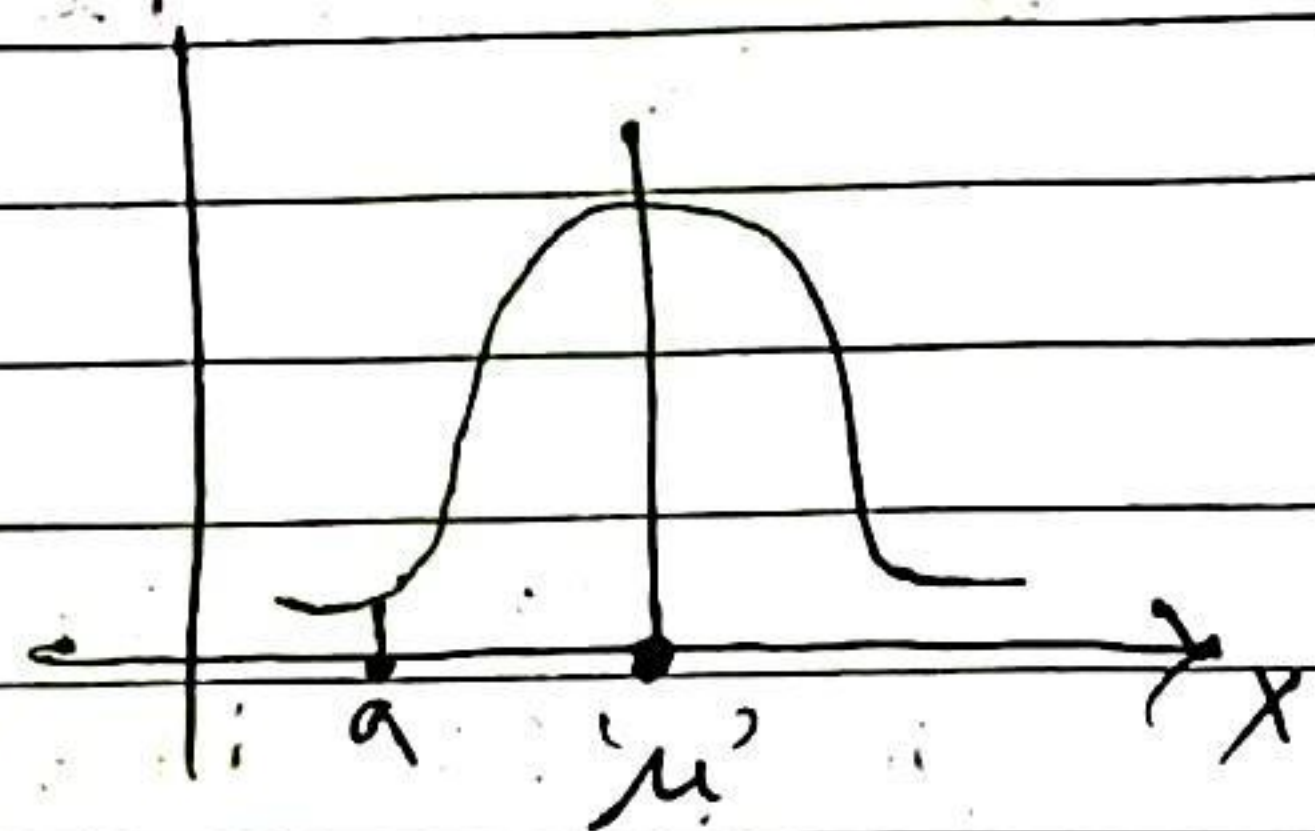
→ we say that a continuous random variable 'x' is a normal random variable (or 'x' is normally distributed) with parameters μ and σ^2 , if the density of 'x' is given by -

$$f(x) = \frac{1}{\sqrt{2\pi} \sigma} e^{-\frac{(x-\mu)^2}{2\sigma^2}} \quad \text{Remember it.}, \quad -\infty < x < \infty.$$

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

(a) $P(X \leq a)$

(b) $P(X > a)$



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

(b) $\int_a^{\infty} f(x) dx$