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~ PART 3 ~

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RANDOM VARIABLES

- # Random Variables: numerical values computed from the outcomes of the experiment.

→ DEFINITION

A random variable is a function with domain as the sample space of an experiment and range as the real numbers, i.e., a function from the sample space to the real line.

- Toss a coin, Sample space = {H, T}
 - Random variable X : $X(H) = 0$, $X(T) = 1$
 - Random variable Y : $Y(H) = -10$, $Y(T) = 200$
 - Random variable Z : $Z(H) = \sqrt{2}$, $Z(T) = \pi$
- While all of the above are valid random variables, usually meaningful functions are considered!

→ EXAMPLE

(1) Throw of Die

$$\text{Sample Space} = \{1, 2, 3, 4, 5, 6\}$$

of random variable X can be defined as follows:

- Pick any 6 real numbers $x_1, x_2, x_3, x_4, x_5, x_6$.
- $X(1) = x_1$, $X(2) = x_2$, $X(3) = x_3$, $X(4) = x_4$, $X(5) = x_5$,
- $X(6) = x_6$

What values for $x_1, x_2, x_3, x_4, x_5, x_6$?

- If x_i 's are distinct, random variable is a one-to-one function
 - Essentially, same as sample space
- However, x_i 's need not be distinct
 - Random variable $E: E(2) = E(4) = E(6) = 1$,
 - $E(1) = E(3) = E(5) = 0$

Random Variable And Events

If X is a random variable,
 $\{x < x\} = \{s \in S : X(s) < x\}$ is an event for all real x .

So, $(x < x)$, $(x = x)$, $(x \leq x)$, $(x \geq x)$ are all events.

Example : Throw a die

$$S = \{1, 2, 3, 4, 5, 6\}$$

$$X(1) = 1, X(2) = 2, X(3) = 3, X(4) = 4, X(5) = 5, X(6) = 6$$

→ $X=1$: die shows 1, $X < 4$: event $\{1, 2, 3\}$

→ Any event can be expressed in terms of X .
 . Event $\{2, 5\}$: $(X=2) \cap (X=5)$

Why Random Variables?

- IPL powerplay over : Several random variables can be defined
 - No. of runs in the over
 - No. of actual deliveries
 - No. of wickets
 - No. of boundaries
 - No. of dot balls , etc.
- Instead of trying to assign probabilities to the entire outcome, we will usually study random variables and assign probabilities to events defined through them.
 - This reduces the detail in the outcome to something simpler.
 - Using limited data, only such random variables can be studied

RANGE OF RANDOM VARIABLE

Range of random variable is the set of values taken by it. Range is a subset of the real line.

- Given the definition of a random variable, the first task is to find its range.

Examples : Throw a die.

- $X = \text{no. shown on a die}$, Range = {1, 2, 3, 4, 5, 6}

$\rightarrow E = 0$ if no. is odd, $E = 1$ if no. is even

$$\text{Range} = \{0, 1\}$$

DISCRETE

DISCRETE RANDOM VARIABLE & DISTRIBUTIONS

Range of a Random Variable

\rightarrow The range of a random variable is the set of values taken by it. Range is the subset of a real line.

* Given the definition of a random variable, the first task is to find its range.

Examples: (1) Throw a die. $X = \text{no. shown on a die}$

$$\text{Range} = \{1, 2, 3, 4, 5, 6\}$$

(2) Throw a die, $E = 0$ (odd), $E = 1$ (even)

$$\text{Range} = \{0, 1\}$$

Discrete Random Variable

\rightarrow A random variable is said to be discrete if its range is a discrete set.

Distribution of a Discrete Random Variable: Probability Mass Function (PMF)

→ PMF : The probability mass function (PMF) of a discrete random variable X with range set T is the function $f_x : T \rightarrow [0, 1]$ defined as

$$f_x(t) = P(X=t) \quad \text{for } t \in T$$

* $(X=t)$ is an event

$P(X=t) = P(\text{all outcomes that result in } X \text{ taking value } t)$

* Let B be a subset of T . Consider the event $(X \in B)$

$$\begin{aligned} P(X \in B) &= P(\text{all outcomes that result in } X \text{ taking values in } B) \\ &= \sum_{t \in B} P(X=t) = \sum_{t \in B} f_x(t) \end{aligned}$$

EXAMPLE

(1) Suppose a fair coin is flipped 3 times.

(a) How many heads will appear?

(b) Which will be the first flip (if any) that shows heads?

Solution: Let $X = \text{no. of heads}$

$\gamma = \text{first flip that shows head}$

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

outcome	$P(\text{outcome})$	X	Y
HHH	$\frac{1}{8}$	3	1
HHT	$\frac{1}{8}$	2	1
HTH	$\frac{1}{8}$	2	1
HTT	$\frac{1}{8}$	1	1
THH	$\frac{1}{8}$	2	2
THT	$\frac{1}{8}$	1	2
TTH	$\frac{1}{8}$	1	3
TTT	$\frac{1}{8}$	0	none

$$X \in \{0, 1, 2, 3\}$$

$$Y \in \{\overset{0}{\text{none}}, 1, 2, 3\}$$

$$f_X(0) = P(TTT) = \frac{1}{8}$$

$$f_Y(0) = \frac{1}{8}$$

$$f_X(1) = P(HHT, HTH, TTH) = \frac{3}{8}$$

$$f_Y(1) = \frac{3}{8}$$

$$f_X(2) = P(HHT, HTH, THH) = \frac{3}{8}$$

$$f_Y(2) = \frac{2}{8}$$

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

$$f_Y(3) = \frac{1}{8}$$

PROPERTIES OF P.M.F

Random Variable, Range and PMF

- Random variable X with range $T = \{t_1, t_2, \dots, t_k\}$,
PMF f_x

t	t_1	t_2	t_3	...	t_k
$f_x(t)$	$f_x(t_1)$	$f_x(t_2)$	$f_x(t_3)$...	$f_x(t_k)$

- PROPERTIES OF PMF

$$\rightarrow 0 \leq f_x(t) \leq 1$$

$$\rightarrow \sum_{t \in T} f_x(t) = 1$$

- Often, random variables and their PMFs are studied without much mention of experiment or sample space

- Events are defined using random variables.
- probabilities evaluated using PMF.

EXAMPLE : Working With PMF 1

- (1) → PMF of a random variable X is given partially in the foll. table :

t	-1	1	2	4
$f_x(t)$	0.5	0.25	0.125	

Find $f_x(4)$. Find the range of X . Find $P(X > 3)$. Find $P(X < \frac{3}{2})$.

Solution: We know,

$$\sum f_x(t) = 1$$

$$\Rightarrow f_x(-1) + f_x(1) + f_x(2) + f_x(4) = 1$$

$$\Rightarrow 0.5 + 0.25 + 0.125 + f_x(4) = 1$$

$$\Rightarrow 0.875 + f_x(4) = 1$$

$$\Rightarrow f_x(4) = 1 - 0.875 = 0.125$$

(b) Range of $X \rightarrow X = \{-1, 1, 2, 4\}$

(c) $P(X > 3) = P(X=4) = f_x(4) = 0.125$

(d) $P(X < \frac{3}{2}) = P(X < 1.5) = P(X=1) + P(X=-1) = 0.25 + 0.5 = 0.75$

(2) The PMF of a random variable X is given below:

$$f_x(k) = \frac{c}{3^k}, \text{ for } k = 1, 2, 3, \dots$$

Find c . Find $P(X > 10)$. Find $P(X > 10 | X > 5)$.

k	1	2	3	...	n
$f_x(k)$	$c/3$	$c/9$	$c/27$...	$c/3^n$

$$\sum f_x(k) = 1 \Rightarrow \frac{c}{3} + \frac{c}{9} + \frac{c}{27} + \dots + \frac{c}{3^n} = 1$$

Formula of Geometric Progression

$$a + a^2 + a^3 + \dots = \frac{a}{1-a} \quad [\text{if } a < 1]$$

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$$\Rightarrow c \left[\frac{1}{3} + \frac{1}{9} + \frac{1}{27} + \dots + \frac{1}{3^n} \right] = 1$$

$$\Rightarrow c \left[\frac{\frac{1}{3}}{1 - \frac{1}{3}} \right] = 1$$

$$\Rightarrow c \left[\frac{1}{2} \right] = 1$$

$$\Rightarrow c = 2$$

Now, we need to find $P(X > 10)$, which means

$$(b) P(X > 10) = P(X=11) + P(X=12) + \dots = \frac{2}{3^{11}} + \frac{2}{3^{12}} + \dots = 2 \left[\frac{1}{3^{11}} + \frac{1}{3^{12}} + \dots \right]$$

$$\Rightarrow P(X > 10) = 2 \left[\frac{\frac{1}{3^{11}}}{1 - \frac{1}{3}} \right] = 2 \left[\frac{1}{3^{11}} \times \frac{3}{2} \right] = \frac{1}{3^{10}}$$

$$(c) P(X > 10 | X > 5) = \frac{P([X > 10] \cap [X > 5])}{P(X > 5)} = \frac{P(X > 10)}{P(X > 5)}$$

$$\therefore P(X > 10 | X > 5) = \frac{\frac{1}{3^{10}}}{\frac{1}{3^5}} = \frac{1}{3^5}$$

COMMON DISTRIBUTIONS

(1) Uniform Random Variable

$x \sim \text{Uniform}(T)$, where T is some finite set

Range : Finite set T

PMF : $f_x(t) = \frac{1}{|T|}$, for all $t \in T$

size of T

Example : (1) Toss a fair coin,

$x \sim \text{Uniform}(\{0, 1\})$, where 0 - head, 1 - tail

(2) Throw a fair die,

$x \sim \text{Uniform}(\{1, 2, 3, 4, 5, 6\})$

(2) Bernoulli Random Variable

$x \sim \text{Bernoulli}(p)$, where $0 \leq p \leq 1$

RANGE : $\{0, 1\}$

PMF : $f_x(0) = 1-p$, $f_x(1) = p$

Example : Bernoulli trial,

p = probability of success, $x \sim \text{Bernoulli}(p)$

(3) Binomial Random Variable

$X \sim \text{Binomial}(n, p)$, where n : the integer, $0 \leq p \leq 1$

RANGE : $\{0, 1, 2, \dots, n\}$

$$\text{PMF} : f_X(k) = {}^n C_k (p)^k (1-p)^{n-k}$$

EXAMPLE :

- No. of successes in n independent Bernoulli (p) trials
- Check the above PMF is valid
 - ${}^n C_k (p)^k (1-p)^{n-k} \geq 0$
 - $\sum_k {}^n C_k (p)^k (1-p)^{n-k} = 1$

(4) Geometric Random Variable

$X \sim \text{Geometric}(p)$, where $0 \leq p \leq 1$

RANGE : $\{1, 2, 3, \dots\}$

$$\text{PMF} : f_X(k) = (1-p)^{k-1} p$$

EXAMPLE :

- No. of trials for first success in repeated independent Bernoulli (p) trials.
- Check that the above PMF is valid
 - Sum of G.P. formula

$$\begin{aligned} a + ar + ar^2 + \dots &= \frac{a}{1-r} \\ a + a^2 + a^3 + \dots &= \frac{a}{1-a} \quad (\text{if } a < 1) \end{aligned}$$

Generalisation of Geometric R.V.

(5) Negative Binomial Random Variable

$X \sim \text{Negative Binomial}(r, p)$, where r : the integer, $0 < p \leq 1$

RANGE : $\{r, r+1, r+2, \dots\}$

$$\text{PMF} : f_X(k) = {}^{k-1}C_{r-1} (1-p)^{k-r} p^r$$

EXAMPLE :

- No. of trials for r success in repeated independent Bernoulli (p) trials
→ $r=1$: geometric random variable
- Check that the above PMF is valid

(6) Poisson Random Variable

$X \sim \text{Poisson}(\lambda)$, where $\lambda > 0$ (real no.)

RANGE : $\{0, 1, 2, 3, \dots\}$

$$\text{PMF} : f_X(k) = \frac{e^{-\lambda} \cdot \lambda^k}{k!}$$

k	0	1	2	3	4	...
$f_X(k)$	$e^{-\lambda}$	$e^{-\lambda} \cdot \lambda$	$\frac{e^{-\lambda} \lambda^2}{2}$	$\frac{e^{-\lambda} \lambda^3}{6}$	$\frac{e^{-\lambda} \lambda^4}{24}$

- Check PMF is valid : use expansion for $e^\lambda = 1 + \lambda + \frac{\lambda^2}{2!} + \frac{\lambda^3}{3!} + \dots$

(7) Hypergeometric Random Variable

$X \sim \text{HyperGeo}(N, r, m)$, where N, r, m : the integers

- Consider a population of ' N ' persons with ' r ' of Type 1 and ' $N-r$ ' of Type 2
- Select ' m ' persons uniformly at random without replacement
- $X = \text{no. of persons of Type 1 selected}$

Range of X

- $N = 100, r = 50, m = 20 \Rightarrow X \in 0, 1, 2, \dots, 20$
- $N = 100, r = 10, m = 20 \Rightarrow X \in 0, 1, 2, \dots, 10$
- $N = 100, r = 90, m = 20 \Rightarrow X \in 10, 11, 12, \dots, 20$

Range

$$X \in \max(0, m - (N-r)), \dots, \min(r, m)$$

PMF of HyperGeo R.V

$$\text{PMF} = f_X(k) = \frac{{}^r C_k {}^{N-r} C_{m-k}}{N C_m}$$

Application Of Poisson Random Var.

Events occurring over a period of time

EXAMPLES:

- Arrival of a person to a booking counter queue
- Arrival of a visitor to a website
- Emission of a particle by a radioactive decay
- of meteorite entering earth's atmosphere.

What is common to above examples?

- Arrival Rate : can be assumed to be constant
 - Given 1 arrival , the time for next arrival can be assumed to be independent.
- * Under above assumptions , the no. of arrivals in a fixed period of time becomes a Poisson R.V.

EXAMPLE 1 : Radioactive decay

In 2608 time intervals of 7.5 seconds each ,
emission of alpha particles is as follows :

particles	→ 0	1	2	3	4	5	6	7	8
times	→ 57	203	383	525	532	408	273	139	45
fraction	→ 0.022	0.078	0.147	0.201	0.204	0.156	0.105	0.053	0.017

$$\text{Emission Rate} = \frac{\text{total no. of particles}}{2608} = 3.8673 (\lambda)$$

Time of next emission : independent of past

$$\text{Poisson Model} : P(\text{no. of particle} = k) = \frac{e^{-\lambda} \lambda^k}{k!}$$

The fraction given above should be closely close to probability under Poisson model.

FUNCTIONS OF ONE RANDOM VAR.

X : a random variable with range T and PMF $f_X(t)$

X : function from sample space to T

T : subset of real line

Let $f(x)$ be a function from real line to real line.

Then, $f(X)$ can be seen as a composition of two functions

so, $f(X)$ is also a random variable in the same probability space.

General Case : X : random variable with PMF $f_X(t)$

$f(X)$: random variable whose PMF is given as follows:

$$f_{f(x)}(a) = P(f(X) = a) = P(X \in \{t : f(t) = a\}) = \sum_{t : f(t) = a} f_X(t)$$