

Exam Roll No. - CSE214021

Class Roll No. - 00191 0501061

BSE 2nd year, 2nd semester, 2020-21, Math-IV, part-2

part-2

1.

a) Chebyshev's inequality :-

If x is a random variable with finite mean μ and variance σ^2 , then, for any value $k > 0$

$$P\{|x - \mu| \geq k\} \leq \frac{\sigma^2}{k^2}$$

b) Bayes' theorem -

Bayes' theorem is stated as the following equation :

$$P(A/B) = \frac{P(B/A) \cdot P(A)}{P(B)}$$

where A and B are events and $P(B) \neq 0$

it may be derived from the definition of conditional probability:

$$\begin{aligned} P(A/B) &= \frac{P(A \cap B)}{P(B)}, \text{ if } P(B) \neq 0 \\ &= \frac{P(B \cap A)}{P(B)} \end{aligned}$$

② general form :- Let A_1, A_2, \dots, A_n be n pairwise mutually exclusive and exhaustive events connected with the random experiment E .

(2)

Let x be an arbitrary event connected with E , where $P(x) \neq 0$. Also, let the probabilities $P(x/A_1), P(x/A_2), \dots, P(x/A_n)$ be all known. Then

$$P(A_i/x) = \frac{P(A_i) \cdot P(x/A_i)}{\sum_{b=1}^n P(A_b) \cdot P(x/A_b)}, \quad i=1, 2, 3, \dots$$

e) we say that x is a uniform random variable on the interval (α, β) if the probability density function of x is given by-

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

2.

a) Number of ways of choosing 3 tickets from 20 tickets (numbered through 1, 2, ..., 20) = ${}^{20}C_3$

Let, x and y be 2 integers

then, arithmetic mean between them (AM) is given by $AM = \frac{x+y}{2}$

\therefore AM must be even. So, summation of 2 integers is even iff both are even or both are odd.

Number of ways of choosing 2 even numbers or two odd numbers from $\{1, 2, \dots, 20\}$

$$= {}^{10}C_2 + {}^{10}C_2 = 2 \cdot {}^{10}C_2$$

$$\therefore \text{Required probability} = \frac{2 \cdot {}^{10}C_2}{{}^{20}C_3} \approx 0.0789$$

b).

Number of white balls = N_1 Number of black balls = N_2

Let number of black balls preceding the first white ball be k and the required probability is $P(k)$

For $k=1$

$$P(1) = \frac{N_2}{N_1 + N_2} \times \frac{N_1}{(N_1 + N_2) - 1}$$

For $k=2$

$$P(2) = (\text{Black}) (\text{Black}) (\text{white})$$

$$= \frac{N_2}{N_1 + N_2} \cdot \frac{N_2 - 1}{N_1 + N_2 - 1} \cdot \frac{N_1}{N_1 + N_2 - 2}$$

For $k=i$

$$P(i) = \frac{N_2}{N_1 + N_2} \cdot \frac{N_2 - 1}{N_1 + N_2 - 1} \cdots \frac{N_2 - i + 1}{N_1 + N_2 - i + 1} \cdot \frac{N_1}{N_1 + N_2 - i}$$

$$= \frac{N_2 c_i}{(N_1 + N_2) c_i} \cdot \frac{N_1}{N_1 + N_2 - i}$$

e) Total number of outcomes possible for (a, b) pair (where $1 \leq a \leq 6 \times 1 \leq b \leq 6$)

$$= 36 \quad (6 \times 6)$$

Now, the required probability is for getting atleast once a double six $(6, 6)$,

It can be found by computing

$$1 - (\text{No. of times no double six occurs})$$

(4)

probability of getting a double 6 (6,6)
 $= \frac{1}{36}$

probability of ~~getting~~ not getting a double

$$\sin(6,6) = 1 - \frac{1}{36} = \frac{35}{36}$$

Total no. of throws = 100

probability of ~~no~~ ^{no} occurrence of a double 6 in 100 throws = $\left(\frac{35}{36}\right)^{100}$

$$\therefore \text{required probability} = 1 - \left(\frac{35}{36}\right)^{100}$$

3)

a)

Joint density of two random variable X and Y are given by,

$$f(x,y) = \begin{cases} e^{-(x+y)} & 0 < x < \infty, 0 < y < \infty \\ 0 & \text{otherwise} \end{cases}$$

Let density function of the random variable

$$\frac{X}{Y} \text{ is } f_{X/Y}(a)$$

$$F_{X/Y}(a) = P\left\{\frac{X}{Y} \leq a\right\}$$

$$= \iint e^{-(x+y)} dx dy$$

$$= \iint_{(x,y): \frac{x}{y} \leq a} e^{-(x+y)} dx dy$$

$$= \int_0^{\infty} \int_0^{ay} e^{-(x+y)} dx dy$$

$$= \int_0^{\infty} \left[\int_0^{ay} e^{-x} dx \right] e^{-y} dy$$

$$= \int_0^{\infty} \frac{e^{-ay}}{e^{-y}} dy = \int_0^{\infty} e^{-(a-1)y} dy$$

$$= \int_0^{\infty} \int_0^{ay} (e^{-x-y}) dx dy$$

$$= \int_0^{\infty} \left[\int_0^{ay} (e^{-x}) dx \right] e^{-y} dy$$

$$= \int_0^{\infty} [e^{-x}]_0^{ay} e^{-y} dy$$

$$= \int_0^{\infty} (1 - e^{-ay}) e^{-y} dy$$

$$= \int_0^{\infty} [e^{-y} - e^{-(a+1)y}] dy$$

$$= \left[-e^{-y} + \frac{e^{-(a+1)y}}{(a+1)} \right]_0^{\infty}$$

$$= [-e^{-y}]_0^{\infty} + \left[\frac{e^{-(a+1)y}}{a+1} \right]_0^{\infty}$$

$$= 1 + \frac{0 - 1}{a+1} = 1 - \frac{1}{a+1}$$

Now, to calculate the density function of ~~x~~ $\frac{X}{Y}$, we differentiate (i), which gives, $f_{\frac{X}{Y}}(a) = \frac{1}{(a+1)^2}$ $0 < a < \infty$

b)

(i) from Markov's Inequality we can say,

If x is a random variable that takes only non-negative values, then, for any value $a > 0$,

$$P\{x \geq a\} \leq \frac{E(x)}{a}$$

Here, number of ~~and~~ item produced during a week is a random variable with mean 50.

* So, probability that this week's production exceeds 75 \Rightarrow

$$P(X \geq 75) \leq \frac{E(X)}{75}$$

\Rightarrow Here $E(X) = 50$

$$\Rightarrow P(X \geq 75) \leq \frac{50}{75} = \frac{2}{3}$$

(ii)

Variance (σ^2) of a week's production is 25.

by Chebyshev's Inequality we can say

If X is a random variable with finite mean μ and variance σ^2 , then for any value $k > 0$

$$P\{|X - \mu| \geq k\} \leq \frac{\sigma^2}{k^2}$$

So, Here $k = 60 - 50 = 50 - 40 = 10$ and $\sigma^2 = 25$ and $\mu = 50$

\therefore probability that this week's production will be between 40 and 60 $\Rightarrow P\{|X - 50| < 10\}$

$$1 - P\{|X - 50| \geq 10\}$$

Here, by Chebyshev's inequality,

$$P\{|X - 50| \geq 10\} \leq \frac{\sigma^2}{10^2} = \frac{25}{100} = \frac{1}{4}$$

$$\therefore P\{|X - 50| < 10\} = 1 - P\{|X - 50| \geq 10\} = 1 - \frac{1}{4} = \frac{3}{4}$$

4.

a)

Chapman-Kolmogorov equation:-

$$P_{ij}^{(n)} = \sum_{k=0}^M P_{ik}^{(b)} P_{kj}^{(n-b)} \quad \text{for all } 0 < b < n$$

Proof:-

$$P_{ij}^{(n)} = P\{X_n = j | X_0 = i\} = \sum_k P\{X_n = j, X_b = k | X_0 = i\}$$

$$= \sum_k P\{X_n = j | X_b = k, X_0 = i\} \cdot P\{X_b = k | X_0 = i\}$$

$$= \sum_k P_{ki}^{(b)} \cdot P_{kj}^{(n-b)} = P_{ij}^{(n)}$$

For a large number of Markov chains it turns

out that $P_{ij}^{(n)}$ converges to a value π_j as $n \rightarrow \infty$.

π_j depends only on j - i.e. for large 'n', the prob. of being in state j , after n transitions is approximately equal to π_j , no matter what the initial state was.

The sufficient condition for a Markov chain to possess the above property is that

for some $n > 0$, $P_{ij}^{(n)} > 0$ for all $i = 0, 1, \dots, M \rightarrow (1)$

Markov chains satisfying (1), is said to be ergodic.

Now, (1) yields

$$P_{ij}^{(n+1)} = \sum_{k=0}^M P_{ik}^{(n)} P_{kj}^{(1)} \rightarrow (2)$$

It follows by letting $n \rightarrow \infty$ for ergodic chains

$$\pi_j = \sum_{k=0}^M \pi_k P_{kj} \rightarrow (3)$$

(8)

Furthermore, since $1 = \sum_{j=0}^M P_{ij}^{(n)}$, we also obtain, by

$$\text{letting } n \rightarrow \infty \quad \sum_{j=0}^M \pi_j = 1 \quad (4)$$

it can be shown that $\pi_j, 0 \leq j \leq M$ are unique non-negative solutions of eqn (3) and (4)

Theorem:

~~For an ergodic Markov~~

Let For an ergodic Markov chain,

$$\pi_j = \lim_{n \rightarrow \infty} P_{ij}^{(n)} \text{ exist and } \pi_j, 0 \leq j \leq M \text{ are}$$

unique non-negative solution of

$$\pi_j = \sum_{k=0}^M \pi_k P_{kj}$$

$$\Rightarrow \sum_{j=0}^M \pi_j = 1$$

from this equation theorem, the limiting probabilities π_0 and π_1 of pain and no pain are given by,

$$\pi_0 = \alpha \pi_0 + \beta \pi_1$$

$$\Rightarrow \pi_1 = (1-\alpha) \pi_0 + (1-\beta) \pi_1$$

$$\Rightarrow \pi_0 + \pi_1 = 1$$

which yields,

$$\pi_0 = \frac{\beta}{1+\beta-\alpha}, \quad \pi_1 = \frac{1-\alpha}{1+\beta-\alpha}$$

~~For~~ ~~no~~ given $\alpha = 0.6$ and $\beta = 0.3$

then the limiting probability of pain on the n th

$$\text{day is } \pi_0 = \frac{0.3}{1+0.3-0.6} = \frac{3}{7}$$

$$\pi_1 = \frac{1-0.6}{1+0.3-0.6} = \frac{4}{7}$$

e)

absorbing and transient states for finite Markov chain :-

A state of a Markov chain is called an absorbing state, if once the Markov chain enters the state, it remains there forever.

i.e., $P_{nn} = 1$ and $P_{nj} = 0$ for $j \neq n$ and $0 \leq n \leq M$

A state is called transient if the system, starts from ~~that~~ that particular state and have zero probability of returning to the same state.