Assignment Ncert Exampler

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I. Question:- GATE 2023.54

Suppose that X is a discrete random variable with the following probability mass

$$P(X=0) = \frac{1}{2} \left(1 + e^{-1} \right) \tag{1}$$

$$P(X = k) = \frac{e^{-1}}{2k!} \text{ for } k = 1, 2, 3, \dots$$
 (2)

Which of the following is/are true?

- 1) E(X) = 1
- 2) E(X) < 1
- 3) $E(X|X > 0) < \frac{1}{2}$ 4) $E(X|X > 0) > \frac{1}{2}$

Solution:

- 1) **Theory:**
 - a) As we know,

$$E(X) = \sum k p_X(k) \tag{3}$$

Therefore,

$$E(X) = 0 \cdot \frac{1}{2} \left(1 + e^{-1} \right) + \sum_{k=1}^{\infty} \frac{k e^{-1}}{2k!}$$
 (4)

$$=\sum_{k=1}^{\infty} \frac{e^{-1}}{2(k-1)!} \tag{5}$$

$$=\frac{1}{2e}\sum_{k=1}^{\infty}\frac{1}{(k-1)!}$$
 (6)

$$= \frac{1}{2e} \cdot e$$
 (Using standard result of exponential series)
$$= \frac{1}{2}$$
 (7)

b) To find E(X|X>0), first we need to find Pr(X|X>0) which can be given as:

$$\Pr(X|X>0) = \frac{\Pr(X=k)}{1 - \Pr(X=0)}$$
(8)

$$=\frac{e^{-1}}{2k!}\cdot\frac{1}{(1-\frac{1}{2}(1+e^{-1}))}\tag{9}$$

$$=\frac{e^{-1}}{2k!}\cdot\frac{2}{(1-e^{-1})}\tag{10}$$

$$=\frac{e^{-1}}{k!(1-e^{-1})}\tag{11}$$

$$=\frac{1}{k!(e-1)}$$
 (12)

Therefore,

$$E(X|X>0) = \sum_{k=1}^{\infty} k \frac{1}{k!(e-1)}$$
 (13)

$$=\frac{1}{e-1}\sum_{k=1}^{\infty}\frac{1}{(k-1)!}$$
(14)

$$= \frac{1}{e-1} \cdot e$$
 (Using standard result of exponential series)
= 1.582 (15)

Referring to equations (7) and (15), we get that option (2) and (4) are correct.

2) Simulation:

To make the simulation of the given question, generate a large set of random variables, say X, with the probability:

$$P(X=0) = \frac{1}{2} \left(1 + e^{-1} \right) \tag{16}$$

$$P(X = k) = \frac{e^{-1}}{2k!} \text{ for } k = 1, 2, 3, \dots.$$
 (17)

To simulate random variables, you kickstart the process by setting the initial conditions for a random number generator, often using the current time to ensure different random sequences each time the program runs. In a loop, you generate a random number between 0 and 1. This whole technique is based on a fascinating concept known as Inverse Transform Sampling. According to this theorem, you have a uniform random variable $U \sim Unif(0, 1)$, and you connect it to the distribution you want using

$$X = F_X^{-1}(U)$$
. (F_X refers to Cumulative Distribution Function)

In simpler terms, you're using U to navigate through the distribution's cumulative probabilities to find the corresponding value of X. It link these random numbers to specific distributions and it relies on a key formula:

$$X = F_X^{-1}(U) = -\frac{ln(1-U)}{\lambda}$$
 (λ depends on probability given)

Using this method, random number is compared to cumulative probabilities (CDF) for various values of therandom variable k until the CDF exceeds it. This process continues until a match is found and k is returned as the generated random variable.

a) To get E(X), take the weighted sum of all possible values of X including 0, each multiplied by its respective probability. Expression for this can be given as:

$$E(X) = \sum k p_X(k) \tag{18}$$

b) As given in theory, to get E(X|X > 0), take the weighted sum of all possible values of X excluding 0, each multiplied by its respective probability and dividing it by the probability of getting random variables greater than 0. Expression for this can be given as:

$$\Pr(X|X>0) = \frac{\Pr(X=k)}{1 - \Pr(X=0)}$$
 (19)

(20)