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AI1103: Assignment 6

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Download Latex codes from

https://github.com/BokkaRajaRaviKiranReddy/ AI1103/blob/main/Assignment6/Assignment6. tex

1 UGC NET 2013 JUNE Q.NO 101

Let $X_1, X_2,...$ be the independent random variables each following exponential distribution with mean 1.then which of the following statements are correct?

- 1) $Pr(X_n > \log n \ \forall n \ge 1)=1$
- 2) $\Pr(X_n > 2 \ \forall n \ge 1) = 1$
- 3) $\Pr(X_n > \frac{1}{2} \ \forall n \ge 1) = 0$
- 4) $\Pr(X_n > \log n, X_{n+1} > \log(n+1) \ \forall n \ge 1) = 0$

2 SOLUTION

Lemma 2.1. PDF of X_i is

$$f_{X_i}(x) = \begin{cases} \lambda_i e^{-\lambda_i x}, & x \ge 0\\ 0, & x < 0 \end{cases}$$

Mean of $X_i = \frac{1}{\lambda_i}$

Proof.

$$E[X_i] = \int_{-\infty}^{\infty} x f_{X_i}(x) dx$$
 (2.0.1)
=
$$\int_{-\infty}^{0} 0 dx + \int_{0}^{\infty} x \lambda_i e^{-\lambda_i x}$$
 (2.0.2)
=
$$\frac{1}{-}$$
 (2.0.3)

Lemma 2.2. CDF of X_i is

$$F_{X_i}(x) = \begin{cases} 1 - e^{-\lambda_i x}, & x \ge 0\\ 0, & x < 0 \end{cases}$$

Proof.

$$F_{X_i}(x) = \int_{-\infty}^{x} f_{X_i}(x) dx$$
 (2.0.4)

For x < 0

$$F_{X_i}(x) = \int_{-\infty}^{x} 0 dx$$
 (2.0.5)

$$= 0$$
 (2.0.6)

For $x \ge 0$

$$F_{X_i}(x) = \int_{-\infty}^{x} f_{X_i}(x) dx$$
 (2.0.7)

$$= \int_{-\infty}^{0} 0 dx + \int_{0}^{x} \lambda_{i} e^{-\lambda_{i} x} dx \qquad (2.0.8)$$

$$=1-e^{-\lambda_i x} \tag{2.0.9}$$

so,

$$F_{X_i}(x) = \begin{cases} 1 - e^{-\lambda_i x}, & x \ge 0\\ 0, & x < 0 \end{cases}$$

From (2.0.3) and $E[X_i] = 1$, we have $\lambda_i = 1 \forall i \ge 1$ Now, for some constant $c \ge 0$

$$\Pr(X_n > c) = 1 - F_{X_n}(c) \tag{2.0.10}$$

$$= e^{-c} (2.0.11)$$

We need Borel-Cantelli Lemmas

Lemma 2.3. Let $A_1, A_2,...$ be a sequence of events in some probability space. The Borel-Cantelli lemma states that, if the sum of the probabilities of the events A_n is finite

$$\sum_{n=1}^{\infty} \Pr(A_n) < \infty \tag{2.0.12}$$

then the probability that infinitely many of them occur is 0

$$\Pr\lim_{n\to\infty}\sup A_n=0\tag{2.0.13}$$

Lemma 2.4. If the events A_n are independent and the sum of the probabilities of the A_n diverges to

infinity, then the probability that infinitely many of them occur is 1. If for independent events $A_1, A_2, ...$

$$\sum_{n=1}^{\infty} \Pr(A_n) = \infty \tag{2.0.14}$$

Then

$$\Pr\left(\lim_{n\to\infty}\sup A_n\right) = 1\tag{2.0.15}$$

1) OPTION 1: We can say the events $X_n > \log n$ are independent $\forall n \geq 1$ as X_n are independent random variable. from (2.0.11)

$$\sum_{n=1}^{\infty} \Pr(X_n > \log n) = \sum_{n=1}^{\infty} e^{-\log n}$$
 (2.0.16)

$$=\sum_{n=1}^{\infty} \frac{1}{n}$$
 (2.0.17)

 $= \infty$ (Cauchy's Criterion) (2.0.18)

Now, from (2.4)

$$\Pr\left(X_n > \log n \ \forall n \ge 1\right) \tag{2.0.19}$$

$$= \Pr\left(\lim_{n \to \infty} \sup X_n > \log n\right) \tag{2.0.20}$$

$$= 1$$
 (2.0.21)

∴ Option 1 is correct.

2) OPTION 2: We can say the events $X_n > 2$ are independent $\forall n \ge 1$ as X_n are independent random variable.

From (2.0.11)

$$\sum_{n=1}^{\infty} \Pr(X_n > 2) = \sum_{n=1}^{\infty} e^{-2}$$
 (2.0.22)

$$= \infty \tag{2.0.23}$$

Now, from (2.4)

$$\Pr(X_n > 2 \ \forall n \ge 1)$$
 (2.0.24)

$$=\Pr\left(\lim_{n\to\infty}\sup X_n>2\right) \tag{2.0.25}$$

$$= 1$$
 (2.0.26)

.. Option 2 is correct.

3) OPTION 3: We can say the events $X_n > \frac{1}{2}$ are independent $\forall n \geq 1$ as X_n are independent random variable.

From (2.0.11)

$$\sum_{n=1}^{\infty} \Pr\left(X_n > \frac{1}{2}\right) = \sum_{n=1}^{\infty} e^{-\frac{1}{2}}$$
 (2.0.27)

$$= \infty \tag{2.0.28}$$

Now, from (2.4)

$$\Pr\left(X_n > \frac{1}{2} \ \forall n \ge 1\right) \tag{2.0.29}$$

$$=\Pr\left(\lim_{n\to\infty}\sup X_n > \frac{1}{2}\right) \tag{2.0.30}$$

$$= 1$$
 (2.0.31)

.. Option 3 is incorrect.

4) OPTION 4:We can say the events $X_n > \log n$ are independent $\forall n \geq 1$ as X_n are independent random variable.

Let the event $X_n > \log n, X_{n+1} > \log(n+1)$ be represented by A_n '

From (2.0.11)

$$\sum_{n=1}^{\infty} \Pr(A_n) \tag{2.0.32}$$

$$= \sum_{n=1}^{\infty} \Pr(X_n > \log n) \Pr(X_{n+1} > \log(n+1))$$
(2.0.33)

$$= \sum_{n=0}^{\infty} e^{-\log n} e^{-\log(n+1)}$$
 (2.0.34)

$$=\sum_{n=1}^{\infty} \frac{1}{n(n+1)}$$
 (2.0.35)

$$=\sum_{n=1}^{\infty} \frac{1}{n} - \frac{1}{n+1}$$
 (2.0.36)

$$= 1$$
 (2.0.37)

Now, from (2.3)

$$\Pr(X_n > \log n, X_{n+1} > \log(n+1) \ \forall n \ge 1)$$
(2.0.38)

$$= \Pr\left(\lim_{n \to \infty} \sup(X_n > \log n, X_{n+1} > \log(n+1))\right)$$
(2.0.39)

$$=0$$
 (2.0.40)

.. Option 4 is correct.

Correct options are Option 1,2,4