

# AI1103 : Assignment 6

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Download Latex codes from

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

*Proof.*

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

For  $x < 0$

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

$$= 0 \quad (2.0.6)$$

For  $x \geq 0$

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

$$= \int_{-\infty}^0 0 dx + \int_0^x \lambda_i e^{-\lambda_i x} dx \quad (2.0.8)$$

$$= 1 - e^{-\lambda_i x} \quad (2.0.9)$$

so,

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

□

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

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

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

We need Borel-Cantelli Lemmas

**Lemma 2.3.** Let  $A_1, A_2, \dots$  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

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

□ then the probability that infinitely many of them occur is 0

$$\Pr \limsup_{n \rightarrow \infty} A_n = 0 \quad (2.0.13)$$

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

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

1)  $\Pr(X_n > \log n \quad \forall n \geq 1) = 1$

2)  $\Pr(X_n > 2 \quad \forall n \geq 1) = 1$

3)  $\Pr(X_n > \frac{1}{2} \quad \forall n \geq 1) = 0$

4)  $\Pr(X_n > \log n, X_{n+1} > \log(n+1) \quad \forall n \geq 1) = 0$

## 2 SOLUTION

**Lemma 2.1.** PDF of  $X_i$  is

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

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

*Proof.*

$$E[X_i] = \int_{-\infty}^{\infty} x f_{X_i}(x) dx \quad (2.0.1)$$

$$= \int_{-\infty}^0 0 dx + \int_0^{\infty} x \lambda_i e^{-\lambda_i x} dx \quad (2.0.2)$$

$$= \frac{1}{\lambda_i} \quad (2.0.3)$$

**Lemma 2.2.** CDF of  $X_i$  is

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

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

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

Then

$$\Pr\left(\limsup_{n \rightarrow \infty} A_n\right) = 1 \quad (2.0.15)$$

- 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)

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

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

$$= \infty \text{ (Cauchy's Criterion)} \quad (2.0.18)$$

Now, from 2.4

$$\Pr(X_n > \log n \quad \forall n \geq 1) \quad (2.0.19)$$

$$= \Pr\left(\limsup_{n \rightarrow \infty} X_n > \log n\right) \quad (2.0.20)$$

$$= 1 \quad (2.0.21)$$

$\therefore$  Option 1 is correct.

- 2) OPTION 2: We can say the events  $X_n > 2$  are independent  $\forall n \geq 1$  as  $X_n$  are independent random variable.  
From (2.0.11)

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

$$= \infty \quad (2.0.23)$$

Now, from 2.4

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

$$= \Pr\left(\limsup_{n \rightarrow \infty} X_n > 2\right) \quad (2.0.25)$$

$$= 1 \quad (2.0.26)$$

$\therefore$  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)

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

$$= \infty \quad (2.0.28)$$

Now, from 2.4

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

$$= \Pr\left(\limsup_{n \rightarrow \infty} X_n > \frac{1}{2}\right) \quad (2.0.30)$$

$$= 1 \quad (2.0.31)$$

$\therefore$  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)

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

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

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

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

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

$$= 1 \quad (2.0.37)$$

Now, from 2.3

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

$$= \Pr\left(\limsup_{n \rightarrow \infty} (X_n > \log n, X_{n+1} > \log(n+1))\right) \quad (2.0.39)$$

$$= 0 \quad (2.0.40)$$

$\therefore$  Option 4 is correct.

Correct options are **Option 1,2,4**