Probability

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Abstract—This book provides solved examples on Probability from IES stats question papers.

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- 2 1 Poisson
- 1.1. Using Central Limit theorem, show that

$$e^{-n} \sum_{k=0}^{n} \frac{n^k}{k!} = \frac{1}{2}$$
 (1.1.1)

Solution:

Definition 1. Let a discrete rv X having poission distribution, then **PMF** is given by

$$f(k; \lambda) = \Pr(X = k) = \frac{\lambda^k e^{-\lambda}}{k!}$$
 (1.1.2)

Let $X_1, X_2, X_3, \dots, X_n$ be the i.i.d rv with $X_i \sim Pois(1)$.

$$E(X_i) = \mu = \lambda = 1$$
 (1.1.3)

$$var(X_i) = \sigma^2 = \lambda = 1 \tag{1.1.4}$$

Let a random variable,

$$X = X_1 + X_2 + \dots + X_n \tag{1.1.6}$$

$$\implies X \sim Pois(n)$$
 (1.1.7)

Theorem 1.1 (Classical central limit theorem). Let X_n be a sequence of independent, identically distributed (i.i.d.) random variables. Assume each X has finite mean, $E(X) = \mu$, and finite variance, $Var(X) = \sigma^2$. Let Z_n be the normalized average of the first n random variables.

$$Z_n = \frac{X - n\mu}{\sqrt{n\sigma^2}} \tag{1.1.8}$$

then Z_n converges in distribution to a standard normal distribution

For X

$$\lambda = n \tag{1.1.9}$$

$$\Pr(X = k) = \frac{n^k e^{-n}}{k!}$$
 (1.1.10)

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Corollary 1.2. By theorem, Z_n converges to standard normal distribution as n goes to infinity i.e CDF of Z_n converges to CDF of standard normal distribution.

$$\Pr\left(Z_n \le x\right) \xrightarrow{n \to \infty} = \Phi(x) \tag{1.1.11}$$

Proof.

$$\implies e^{-n} \sum_{k=0}^{n} \frac{n^{k}}{k!} = \sum_{k=0}^{n} e^{-n} \frac{n^{k}}{k!}$$

$$= \sum_{k=0}^{n} \Pr(X = k)$$

$$= \Pr(X \le n)$$

$$= \Pr\left(\frac{X - n\mu}{\sqrt{n\sigma^{2}}} \le \frac{n - n\mu}{\sqrt{n\sigma^{2}}}\right)$$

$$= \Pr(Z_{n} \le 0)$$

$$(1.1.15)$$

Using (1.1.11),

$$\Pr\left(Z_n \le 0\right) \xrightarrow{n \to \infty} \Phi(0) = \frac{1}{2} \tag{1.1.17}$$

2 2015

2.1. For random variables X and Y, show that: Var[Y] = E[Var(Y|X)] + Var[E(Y|X)] Solution:

Let the abbreviations LE and LIE denote linearity of expectations and law of iterated expectations respectively.

$$Var[Y] = E[Y^2] - [E(Y)]^2$$
 (definition)
= $E[E(Y^2|X)] - (E[E(Y|X)])^2$ (LIE)
(2.1.2)

$$= E[E(Y^2|X)] - (E[E(Y|X)])^2$$
$$- E([E(Y|X)]^2) + E([E(Y|X)]^2) \quad (2.1.3)$$

$$= E[E(Y^{2}|X)] - E([E(Y|X)]^{2}) + E([E(Y|X)]^{2}) - (E[E(Y|X)])^{2} \text{ (LE \& LIE)}$$
(2.1.4)

$$= Var[E(Y|X)] + E[Var(Y|X)]$$
 (definition) (2.1.5)

2.2. Let X be a Random Variable with E[X] = 3, $E[X^2] = 13$. Use Chebyshev's Inequality to obtain Pr(-2 < X < 8)

Solution: Let X be a random variable with finite expected value E[X] and finite non-zero variance σ^2 . Then for any real number k > 0,

$$\Pr(|X - E[X]| \ge k\sigma) \le \frac{1}{k^2} \tag{1}$$

computing the variance,

$$\sigma^{2} = E[X^{2}] - E[X]^{2}$$

$$\Rightarrow \sigma^{2} = 13 - 9 = 4 \qquad (2)$$

$$\sigma = 2 \qquad (3)$$

using (3),

$$Pr(-2 < X < 8) = 1 - Pr(|X - 3| > 5)$$
 (4)

$$Pr(|X - 3| > 5) = Pr(|X - E[X]| > k\sigma) \quad (5)$$

$$k\sigma = 5$$

$$\implies 2k = 5$$

$$\therefore k = \frac{5}{2} \quad (6)$$

Using (1), (5) and (6) in (4),

$$\Pr(-2 < X < 8) \ge 1 - \left(\frac{2}{5}\right)^2$$

$$\implies \Pr(-2 < X < 8) \ge \frac{21}{25} \tag{7}$$

2.3. Three points are chosen on the line of unit length. Find the probability that each the 3 line segments have length greater than $\frac{1}{4}$.

Solution:

Let $X, Y \in \{0, 1\}$ be the random variables which represent the position of two points on the line of unit length.

Conditions which should be satisfied to have three line segments with length greater than $\frac{1}{4}$ are given in the below table.

Then the required event which solves the prob-

Hence, proved.

Event	Condition
A	$\frac{1}{4} < X < \frac{3}{4}$
В	$\frac{1}{4} < Y < \frac{3}{4}$
С	$\frac{1}{4} < X - Y$
D	$\frac{1}{4} < Y - X$

TABLE 2.3.1: Events and their conditions

lem is ABC+ABD.

$$\Pr(ABC) = \Pr\left(\frac{1}{4} + Y < X, \frac{1}{4} < X, Y < \frac{3}{4}\right)$$

$$= \sum \Pr\left(Y = y | \frac{1}{4} < X, Y < \frac{3}{4}\right) \times$$

$$\Pr\left(\frac{1}{4} + y < X, \frac{1}{4} < X < \frac{3}{4}\right) \times$$

$$= \int_{\frac{1}{4}}^{\frac{3}{4}} dy f_{Y}(y) \times$$

$$\Pr\left(\frac{1}{4} + y < X, \frac{1}{4} < X < \frac{3}{4}\right)$$

$$(2.3.2)$$

$$= \int_{\frac{1}{4}}^{\frac{3}{4}} dy f_{Y}(y) \Pr\left(\frac{1}{4} + y < X < \frac{3}{4}\right)$$

$$(2.3.3)$$

As X is distributed uniformly between 0 and 1.

$$\Pr\left(\frac{1}{4} + y < X < \frac{3}{4}\right) = \begin{cases} \frac{1}{2} - y & y \in \left(0, \frac{1}{2}\right) \\ 0 & \text{otherwise} \end{cases}$$
(2.3.5)

Using (2.3.5),(2.3.4) can be written as

$$\Pr(ABC) = \int_{\frac{1}{4}}^{\frac{1}{2}} dy f_Y(y) \left(\frac{1}{2} - y\right)$$
 (2.3.6)

As y is distributed uniformly between 0 and 1.

$$\Pr(ABC) = \int_{\frac{1}{4}}^{\frac{1}{2}} \frac{1}{2} - y \, dy \qquad (2.3.7)$$
$$= \frac{1}{32} \qquad (2.3.8)$$

Similarly, we can find,

$$\Pr(ABD) = \frac{1}{32} \tag{2.3.9}$$

As C and D are mutually exclusive events.

$$Pr(ABC + ABD) = Pr(ABC) + Pr(ABD)$$
(2.3.10)

$$=\frac{1}{16}\tag{2.3.11}$$

: probability that each of the three line segments have length greater than $\frac{1}{4}$ is $\frac{1}{16}$.

2.4. Two points are chosen on a line of unit length. Find the probability that each of the 3 line segments will have length greater than $\frac{1}{4}$? **Solution:** let us choose points X,Y on a line ,since we are picking points randomly they have a uniform distribution in (0, 1).

> let us take two random variables X,Y. This points divide in into segments X, Y - X, 1 - XFrom question

$$X > \frac{1}{4} \tag{2.4.1}$$

$$X > \frac{1}{4}$$
 (2.4.1)
 $Y - X > \frac{1}{4}$ (2.4.2)

$$1 - X < 1/4 \tag{2.4.3}$$

 $\binom{2}{1} \times \Pr\left(X + \frac{1}{4} < Y < \frac{3}{4}, \frac{1}{4} < X < \frac{1}{2}\right)$ is the required answer since given random variables

X,Y are interchangable.

$$\Pr\left(X + \frac{1}{4} < Y < \frac{3}{4}, \frac{1}{4} < X < \frac{1}{2}\right) \tag{2.4.4}$$

$$= \int_{X=\frac{1}{4}}^{X=\frac{1}{2}} \int_{Y=X+\frac{1}{4}}^{Y=\frac{3}{4}} dY dX \qquad (2.4.5)$$

$$= \int_{X=\frac{1}{4}}^{X=\frac{1}{2}} \left(\frac{1}{2} - X\right) dX \qquad (2.4.6)$$

$$= \left(\frac{X}{2} - X^2\right) \Big|_{\frac{1}{2}}^{\frac{1}{2}} \qquad (2.4.7)$$

$$= \frac{1}{8} - \frac{1}{8} + \frac{1}{32} \qquad (2.4.8)$$

$$=\frac{1}{32}$$
 (2.4.9)

So,req answer is

$$2 \times \Pr\left(X + \frac{1}{4} < Y < \frac{3}{4}, \frac{1}{4} < X < \frac{1}{2}\right) = \frac{1}{16}$$

Ans is $\frac{1}{16}$

3 2016

- 3.1. Let the random variable X have the distribution P(X = 0) = P(X = 3) = p, P(X = 1) = 1 3p for $0 \le p \le \frac{1}{2}$. What is the maximum value of V(X)?
 - A) 3
 - B) 4
 - C) 5
 - D) 6
 - E) none

Solution: Given, for $0 \le p \le \frac{1}{2}$,

$$P(X=0) = p (3.1.1)$$

$$P(X=1) = 1 - 3p \tag{3.1.2}$$

$$P(X=3) = p (3.1.3)$$

Now consider P(X = 1) = 1 - 3p for $p = \frac{1}{2}$. We get,

$$P(X=1) = 1 - 3p \tag{3.1.4}$$

$$= 1 - (3)\left(\frac{1}{2}\right) \tag{3.1.5}$$

$$=1-\frac{3}{2} \tag{3.1.6}$$

$$= -\frac{1}{2} < 0 \tag{3.1.7}$$

Probability cannot be negative. But in equation (0.0.7) probability is negative, which is not possible.

Therefore, the question is not a proper one.

Answer: Option E

- 3.2. X_1 and X_2 are independent Poisson variables such that $Pr(X_1 = 2) = Pr(X_1 = 1)$ and $Pr(X_2 = 2) = Pr(X_2 = 3)$. What is the variance of $(X_1 2X_2)$?
 - a) 14
 - b) 4
 - c) 3
 - d) 2

Solution: For a Poisson variable X,

$$\Pr(X = k) = \frac{\lambda^k e^{-\lambda}}{k!}$$
 (3.2.1)

Since $Pr(X_1 = 2) = Pr(X_1 = 1)$,

$$\frac{{\lambda_1}^2 e^{-\lambda_1}}{2!} = \frac{{\lambda_1}^1 e^{-\lambda_1}}{1!}$$
 (3.2.2)

$$\lambda_1 = 2!/1! = 2 \tag{3.2.3}$$

Similarly, as $Pr(X_2 = 2) = Pr(X_2 = 3)$,

$$\frac{{\lambda_2}^2 e^{-\lambda_2}}{2!} = \frac{{\lambda_2}^3 e^{-\lambda_2}}{3!}$$
 (3.2.4)

$$\lambda_2 = 3!/2! = 3$$
 (3.2.5)

Also we know for a Poisson variable X, the following holds true:

$$E[X] = \lambda \tag{3.2.6}$$

$$Var[X] = \lambda \tag{3.2.7}$$

$$Var[X] = E[X^2] - (E[X])^2$$
 (3.2.8)

Now, for the variance of $(X_1 - 2X_2)$

$$Var[X_1 - 2X_2] = E[(X_1 - 2X_2)^2] - (E[X_1 - 2X_2])^2$$

$$= E[X_1^2 + 4X_2^2 - 4X_1X_2]$$

$$- (E[X_1] - 2E[X_2])^2$$

$$= E[X_1^2] - (E[X_1])^2 + 4E[X_2^2]$$

$$- 4(E[X_2])^2) + 4E[X_1X_2]$$

$$+ 4E[X_1]E[X_2]$$
 (3.2.9)

Since the variables are independent:

$$E[X_1X_2] = E[X_1]E[X_2]$$
 (3.2.10)

Substituting equations (3.2.7) and (3.2.8), we

get:

$$Var[X_1 - 2X_2] = Var[X_1] + 4(Var[X_2])$$

$$- 4E[X_1][X_2] + 4E[X_1][X_2]$$

$$= \lambda_1 + 4\lambda_2 = 2 + 4(3) = 14$$
(3.2.11)

Hence option (a) 14 is correct.