

Chapter One
Random Variables

Space Sample : The Sample denoted by S is the collection or totality of all possible outcomes of a conceptual experiment.

Random Variable : The random variable denoted by X or $X(\cdot)$ is a real number which associated outcomes of the random experiment. (i.e. The random variable is a function with domain S (sample space) and codomain the real line R , that is a real function defined on the sample space). $p(\text{Random Variable}) : \mathbb{R} \rightarrow [0, 1]$

Discrete Random Variable: A r.v. X will be defined to be discrete if the range of X is countable, say x_1, x_2, x_3, \dots

Continuous Random Variable : A r.v. X will be defined to the continuous if the range of X is uncountable, say $a \leq x_i \leq b \quad i=1, 2, 3, \dots$

Note: The space of a r.v. X is the real numbers $= \{x : X(\omega) = x, \omega \in S\}$

Ex.1: Consider the experiment of tossing a single coin, then the sample space $S = \{\text{head, tail}\}$. Let the r.v. X denote the number of heads, hence

$$X(\omega) = \begin{cases} 0, & \text{if } \omega = \text{tail} \\ 1, & \text{if } \omega = \text{head} \end{cases}$$

$$X(\omega) = \begin{cases} 0, & \text{if } \omega = \text{head} \\ 1, & \text{if } \omega = \text{tail} \end{cases} \leftarrow \text{number of tails}$$

So, the r.v. X associates a real number with each outcomes of the experiment (i.e. $X = \{0, 1\}$)

Ex.2: Consider experiment of tossing two dice, then the sample space $S = \{(i, j) : i, j = 1, 2, \dots, 6\}$. Let the r.v. X denote the sum of the upturned faces, so $X(\omega) = i + j, \omega = (i, j) \in S$; so $X = \{2, 3, 4, \dots, 12\}$.

Probability Density Function : Any function $f(x)$ with domain the real line R and codomain $[0, \infty)$ (i.e. $f : R \rightarrow [0, \infty)$) is called probability density function (p.d.f.) of a r.v. X if the following conditions are satisfies ;

1. $f(x) \geq 0, \forall x$

2. a) $\sum_x f(x) = 1$, if X a discrete r.v. b) $\int_x f(x) dx = 1$, if X a continuous r.v.

Remark : The probability $\Pr\{X \in A\}$ is completely determined by the probability density function $f(x)$ of the random variable X as following :

a) $\sum_{x \in A} f(x) = \Pr\{X \in A\}$, if X is discrete r.v.

b) $\int_A f(x) dx = \Pr\{X \in A\}$, if X is continuous r.v.

Ex.3: Let the r.v. X have the p.d.f.

$$f(x) = \begin{cases} 2c / (x-1)! & , x = 1, 2, 3, \dots \\ 0 & \text{o.w.} \end{cases}$$

Find 1) The constant c 2) The $\Pr(A)$ where $A = \{x : 2 \leq x \leq 5\}$

Solution:

1. by properties of the p.d.f. $f(x)$ of a r.v. X we get; *مجموعه كثرة غير متناهية*

$$\sum_{x=1}^{\infty} f(x) = 1 \Rightarrow \sum_{x=1}^{\infty} \frac{2c}{(x-1)!} = 1$$

$$\Rightarrow 2c \left(\frac{1}{0!} + \frac{1}{1!} + \frac{1}{2!} + \dots \right) = 1$$

$$\Rightarrow 2c(e^1) = 1 \quad \left(\text{since } e^x = \sum_{i=0}^{\infty} \frac{x^i}{i!} = \frac{x^0}{0!} + \frac{x^1}{1!} + \dots \right)$$

$$\Rightarrow c = \frac{1}{2e} \Rightarrow c = 0.367 \quad \boxed{c = 2.7182}$$

Handwritten notes: $c \times c^2 + c^3 + \dots$, $\sqrt{\frac{c}{1-c}} = \dots$, $\frac{c(1-c^n)}{1-c} = \dots$

2) $\Pr(A) = \Pr(2 \leq X \leq 5) = \sum_{x=2}^5 f(x) = \frac{1}{e} \sum_{x=2}^5 \frac{1}{(x-1)!} = \frac{1}{e} \left(\frac{1}{1!} + \frac{1}{2!} + \frac{1}{3!} + \frac{1}{4!} \right) = 0.63$

Handwritten note: $e^x = \sum_{i=0}^{\infty} \frac{x^i}{i!} = \frac{x^0}{0!} + \frac{x^1}{1!} + \frac{x^2}{2!} + \dots$

Ex.4: Let the r.v. X have the p.d.f. $f(x) = \begin{cases} kxe^{-x} & , x \geq 0 \\ 0 & \text{o.w.} \end{cases}$

find the $\Pr(A_1 \cup A_2)$, $\Pr(A_1 \cap A_2)$ where $A_1 = \{x : 1 \leq x \leq 3\}$ and $A_2 = \{x : 2 \leq x \leq 5\}$

Solution:

Firstly we find the constant k by using the properties of p.d.f. $f(x)$

$$\int_{-\infty}^{\infty} f(x) dx = \int_0^{\infty} kxe^{-x} dx = 1$$

By part integration method $\int u dv = uv - \int v du \Rightarrow \int_0^{\infty} xe^{-x} dx = -xe^{-x} + \int_0^{\infty} e^{-x} dx$

$$= [-xe^{-x} - e^{-x}]_0^{\infty} = [0 + 1] = 1$$

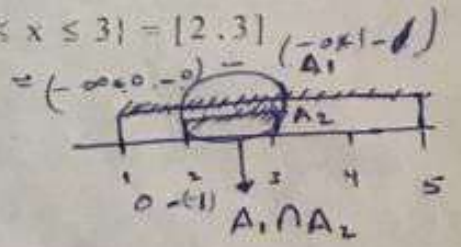
$$\int_0^{\infty} f(x) dx = k \int_0^{\infty} xe^{-x} dx = 1 \Rightarrow k \left([-xe^{-x}]_0^{\infty} + \int_0^{\infty} e^{-x} dx \right) = 1$$

$$\Rightarrow -k[e^{-x}]_0^{\infty} = 1 \Rightarrow k = 1$$

Now $A_1 \cup A_2 = \{x : 1 \leq x \leq 5\} = [1, 5]$ $A_1 \cap A_2 = \{x : 2 \leq x \leq 3\} = [2, 3]$

$$\Pr(A_1 \cup A_2) = \int_1^5 f(x) dx = \int_1^5 xe^{-x} dx = -xe^{-x} \Big|_1^5 + \int_1^5 e^{-x} dx$$

$$= -5e^{-5} + e^{-1} + [-e^{-x}]_1^5 = 2e^{-1} - 6e^{-5} = 0.74 - 0.04 = 0.69$$

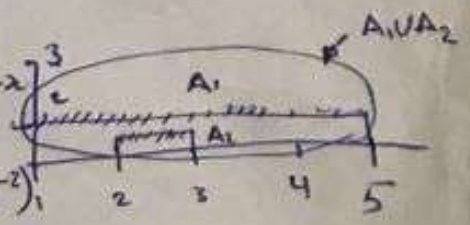


$$\Pr(A_1 \cap A_2) = \int_2^3 f(x) dx = \int_2^3 xe^{-x} dx$$

$$= -xe^{-x} \Big|_2^3 + \int_2^3 e^{-x} dx = -xe^{-x} \Big|_2^3 + [-e^{-x}]_2^3$$

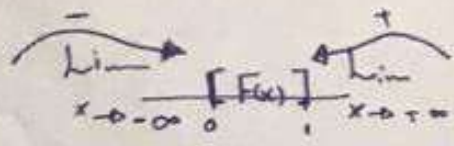
$$= (-3e^{-3} + 2e^{-2}) + (-e^{-3} + e^{-2}) = 2e^{-2} - 4e^{-3}$$

$$= 0.27 - 0.199 = 0.07$$



Cumulative Distribution Function : The cumulative distribution function of a r.v. denoted by $F_X(x)$ or $F(x)$, is defined to be that function with domain the real line \mathbb{R} and codomain the closed interval $[0, 1]$ (i.e. $f: \mathbb{R} \rightarrow [0, 1]$) which satisfies

$$F(x) = \Pr(X \leq x) = \begin{cases} \sum_{u \leq x} f(u) & \text{if } X \text{ a discrete r.v.} \\ \int_{-\infty}^x f(u) du & \text{if } X \text{ a continuous r.v.} \end{cases}$$



Properties of a cumulative distribution function :

- 1) $0 \leq F(x) \leq 1$
- 2) $F(-\infty) = \lim_{x \rightarrow -\infty} F(x) = 0$, $F(+\infty) = \lim_{x \rightarrow +\infty} F(x) = 1$
- 3) $F(x)$ is **monotone, non-decreasing function**; that is if $a < b \Rightarrow F(a) \leq F(b)$
- 4) $F(x)$ is **right continuous**; that is $\lim_{0 < h \rightarrow 0} F(x+h) = F(x)$
- 5) $\Pr(a < X \leq b) = F(b) - F(a)$, $\forall a < b$
- 6) $\Pr(X = x) = F(x) - F(x^-)$

$f(x) = \frac{dF(x)}{dx}$

where $F(x^-)$ are the left hand limit of $F(x)$.

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the p.d.f function is the partial of c.d.f. for the random variable x.

Ex.5: Let $f(x) = \begin{cases} x/15 & x = 1, 2, 3, \dots \\ 0 & \text{o.w.} \end{cases}$ be the p.d.f. of a r.v. X

- 1) Sketch graph of c.d.f. $F(x)$.
- 2) Compute $\Pr(2 \leq X \leq 5)$ by using two methods. *probability? Distribution function?*
- 3) Compute $\Pr(X=2)$ by using two methods. *Same?*

Solution:

1) $F(x) = \Pr(X \leq x) = \sum_{u=1}^x f(u) = \sum_{u=1}^x \frac{u}{15} = \frac{1}{15}(1+2+\dots+x)$

$\sum_{i=1}^x i \rightarrow$ مجموعها $[\frac{x}{2}(x+1)]$

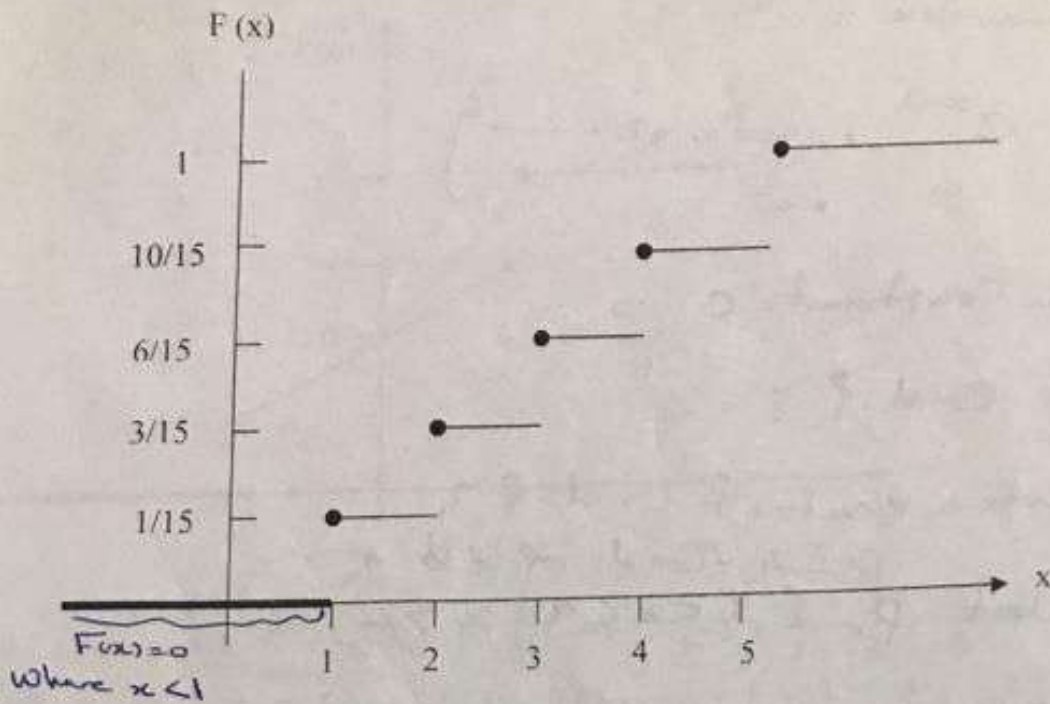
$= \frac{1}{15} [\frac{x}{2}(x+1)]$ since $\sum_{i=1}^x i = 1+2+\dots+x$ is a numerical series

$= \frac{1}{30} x(x+1)$ $x = 1, 2, 3, \dots$

$F(x) = \begin{cases} 0 \\ 1/15 \\ 3/15 \\ 6/15 \\ 10/15 \\ 1 \end{cases}$	$x < 1$	0	$p(x=1) = \frac{1}{15}$
	$1 \leq x < 2$	$1/15$	$p(x=2) = p(x=2) + p(x=1) = \frac{2}{15} + \frac{1}{15} = \frac{3}{15}$
	$2 \leq x < 3$	$3/15$	$p(x=3) = p(x=3) + p(x=2) = \frac{3}{15} + \frac{3}{15} = \frac{6}{15}$
	$3 \leq x < 4$	$6/15$	$p(x=4) = p(x=4) + p(x=3) = \frac{4}{15} + \frac{6}{15} = \frac{10}{15}$
	$4 \leq x < 5$	$10/15$	$p(x=5) = p(x=5) + p(x=4) = \frac{5}{15} + \frac{10}{15} = \frac{15}{15} = 1$
$5 \leq x$	1	1	



4)

2) a. By depending on c.d.f. $F(x)$

$$\Pr(2 \leq X \leq 5) = \Pr(X \leq 5) - \Pr(X \leq 2) = F(5) - F(2^-) = 1 - \frac{1}{15} = \frac{14}{15}$$

b. By depending on p.d.f. $f(x)$

$$\Pr(2 \leq X \leq 5) = \sum_{x=2}^5 f(x) = f(2) + f(3) + f(4) + f(5) = \frac{2}{15} + \frac{3}{15} + \frac{4}{15} + \frac{5}{15} = \frac{14}{15}$$

$$3) \text{ a. } \Pr(X=2) = F(2) - F(2^-) = \frac{3}{15} - \frac{1}{15} = \frac{2}{15} \quad (\text{by depending on c.d.f. } F(x))$$

$$\text{b. } \Pr(X=2) = \sum_{x=2}^2 f(x) = f(2) = \frac{2}{15} \quad (\text{by depending on p.d.f. } f(x))$$

Exercise: Let $f(x) = (c/3)^x$, $x = 1, 2, \dots$ be the p.d.f. of a r.v. X , find the value of c and sketch graph of c.d.f. $F(x)$?

Ex.6: Let $F(x) = \begin{cases} 0 & , x < -1 \\ (x+2)/4 & , -1 \leq x < 1 \\ 1 & , 1 \leq x \end{cases}$ be the c.d.f. of a r.v. X ,

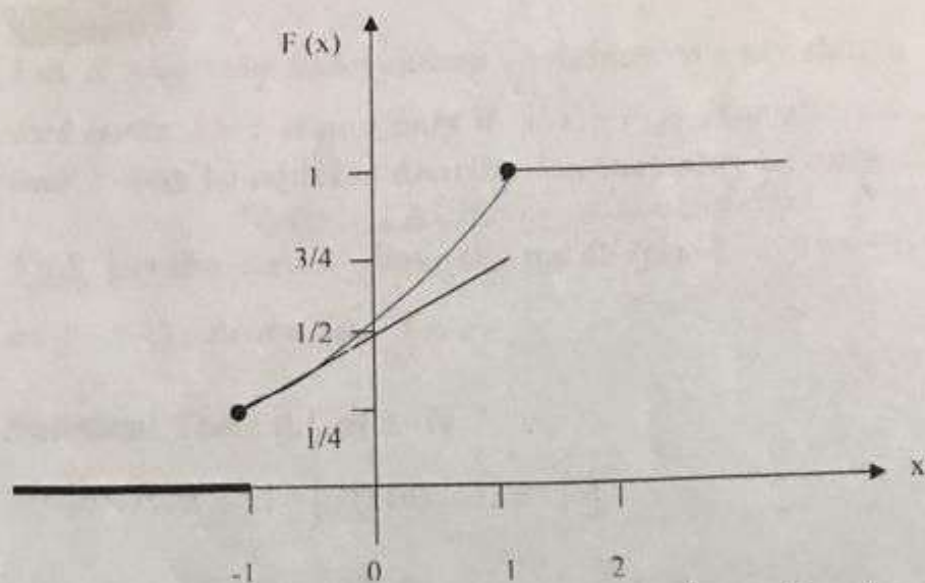
1) Sketch graph of c.d.f. $F(x)$.

2) Compute

$$1) \Pr(-\frac{1}{2} \leq X \leq \frac{1}{2}) \quad 2) \Pr(2 < X \leq 3) \quad 3) \Pr(X = 0) \quad 4) \Pr(X = 1)$$

Solution:

5)



$$1) \Pr\left(-\frac{1}{2} \leq X \leq \frac{1}{2}\right) = F\left(\frac{1}{2}\right) - F\left(-\frac{1}{2}\right) = \frac{5}{8} - \frac{3}{8} = \frac{1}{4}$$

$$2) \Pr(2 < X \leq 3) = F(3) - F(2) = 1 - 1 = 0$$

$$3) \Pr(X = 0) = F(0) - F(0^-) = \frac{2}{4} - \frac{2}{4} = 0$$

$$4) \Pr(X = 1) = F(1) - F(1^-) = 1 - \frac{3}{4} = \frac{1}{4}$$

W **Remarks:**

1) c.d.f. $F(x)$ in above example (Ex.6) is a mixture of the continuous and discrete types.

2) If X is continuous r.v., then p.d.f. $f(x)$ can be obtained by derivative of c.d.f.

$F(x)$, that is $f(x) = \frac{dF(x)}{dx}$

W **Ex.7:** Let $F(x) = \begin{cases} 0 & , & x < 0 \\ \sqrt{x}/2 & , & 0 \leq x < 4 \\ 1 & , & 4 \leq x \end{cases}$ be the c.d.f. of a r.v. X , find the p.d.f. of X

Solution:

Since X is continuous r.v.

Hence

$$f(x) = \frac{dF(x)}{dx} = \frac{1}{4\sqrt{x}}, \quad 0 < x < 4$$

$$\Rightarrow f(x) = \begin{cases} \frac{1}{4\sqrt{x}} & , & 0 < x < 4 \\ 0 & , & \text{o.w.} \end{cases}$$

Remark:

Let X and Y be two random variables. We say that X and Y are equal in distribution and write $X \stackrel{D}{=} Y$ if and only if $F(x) = F(y)$, for all $x \in R$. It is important to note while X and Y may be equal in distribution they may be quite different.

Ex.8 Let the r.v. X have the p.d.f. $f(x) = 1$, $0 < x < 1$. Define the random variable Y as $Y = 1 - X$. Prove that $X \stackrel{D}{=} Y$.

Solution: The c.d.f. of X is

$$F_X(x) = \Pr(X \leq x) = \int_0^x f(u) du = \int_0^x 1 du = [u]_0^x = x$$

$$\Rightarrow F_X(x) = \begin{cases} 0 & , x < 0 \\ x & , 0 \leq x < 1 \\ 1 & , 1 \leq x \end{cases}$$

The space of Y is the interval $(0, 1)$, the same as X (since $Y = 1 - X$). Further, the c.d.f. of Y is

$$F_Y(y) = \Pr(Y \leq y) = \Pr(1 - X \leq y) = \Pr(X \geq 1 - y) = 1 - \Pr(X \leq 1 - y) = 1 - F_X(1 - y) = 1 - (1 - y) = y$$

$$\Rightarrow F_Y(y) = \begin{cases} 0 & , y < 0 \\ y & , 0 \leq y < 1 \\ 1 & , 1 \leq y \end{cases}$$

Hence, Y has the same c.d.f. as X , i.e., $X \stackrel{D}{=} Y$, but $X \neq Y$.

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Expectation: The expectation or expected value of the function $g(X)$ of the r.v. X , denoted by $E(g(X))$ or $\mu_{g(X)}$ is defined by ;

$$E(g(X)) = \begin{cases} \sum_{x_i} g(x_i) f(x_i) & , \text{ if } X \text{ a discrete r.v.} \\ \int_{-\infty}^{\infty} g(x) f(x) dx & , \text{ if } X \text{ a continuous r.v.} \end{cases}$$

Where $f(x)$ is the p.d.f. of X .

Properties of Expected Value

1. $E(c) = c$ for a constant c .
2. $E(c g(X)) = c E(g(X))$ for a constant c .
3. $E(\sum_{i=1}^n c_i g_i(X)) = \sum_{i=1}^n c_i E(g_i(X))$
4. If $g_1(X) \leq g_2(X)$ then $E(g_1(X)) \leq E(g_2(X))$.

Proof:

1. Assume X is a continuous r.v.

$$E(c) = \int_{-\infty}^{\infty} c f(x) dx = c \int_{-\infty}^{\infty} f(x) dx = c \quad (\text{Since } f(x) \text{ is p.d.f. (i.e. } \int_{-\infty}^{\infty} f(x) dx = 1))$$

7)

3. Assume X is continuous r.v.

$$\begin{aligned} E\left(\sum_{i=1}^n c_i g_i(X)\right) &= E(c_1 g_1(X) + c_2 g_2(X) + \dots + c_n g_n(X)) \\ &= \int_{-\infty}^{\infty} (c_1 g_1(x) + c_2 g_2(x) + \dots + c_n g_n(x)) f(x) dx \\ &= c_1 \int_{-\infty}^{\infty} g_1(x) f(x) dx + c_2 \int_{-\infty}^{\infty} g_2(x) f(x) dx + \dots + c_n \int_{-\infty}^{\infty} g_n(x) f(x) dx \\ &= c_1 E(g_1(X)) + c_2 E(g_2(X)) + \dots + c_n E(g_n(X)) = \sum_{i=1}^n c_i E(g_i(X)) \end{aligned}$$

Remarks:

- 1) If $E(X) = \mu_X \rightarrow \infty$, then we say that the mean of a r.v. X does not exist.
 2) The r^{th} moments of the r.v. X , denoted by $E(X^r)$ or μ'_X is defined by:

$$E(X^r) = \begin{cases} \sum_{x=x} x^r f(x) & , \text{ if } X \text{ a discrete r.v.} \\ \int_{-\infty}^{\infty} x^r f(x) dx & , \text{ if } X \text{ a continuous r.v.} \end{cases}$$

where $f(x)$ is the p.d.f. of X .

- 3) The r^{th} central moment of the r.v. X about a is defined as $E((X - a)^r)$

Ex.9: Let $f(x) = 1/x^2$, $x \geq 1$ be the p.d.f. of a r.v. X , does $E(X)$?

Solution:

Since

$$E(X) = \int_1^{\infty} x f(x) dx = \int_1^{\infty} \frac{dx}{x} = [L n(x)]_1^{\infty} = \lim_{x \rightarrow \infty} L n(x) - L n(1) = \infty$$

Hence $E(X)$ does not exist.

Ex.10: Let X be a r.v. with p.d.f. given by $f(x, \lambda) = \lambda e^{-\lambda x}$, $\lambda, x > 0$ (which is called the exponential distribution), find $E(X)$, $E(X^2)$

Solution:

$$E(X) = \int_0^{\infty} x f(x) dx = \int_0^{\infty} \lambda x e^{-\lambda x} dx$$

By part integration method

$$u = x \Rightarrow du = dx$$

$$dv = \lambda e^{-\lambda x} dx \Rightarrow v = -e^{-\lambda x}$$

$$E(X) = [-x e^{-\lambda x}]_{x=0}^{\infty} + \int_0^{\infty} e^{-\lambda x} dx = -\frac{1}{\lambda} [e^{-\lambda x}]_{x=0}^{\infty} = -\frac{1}{\lambda} (0 - 1) = \frac{1}{\lambda}$$

$$E(X^2) = \int_0^{\infty} x^2 f(x) dx = \int_0^{\infty} \lambda x^2 e^{-\lambda x} dx$$

By part integration method

$$u = x^2 \Rightarrow du = 2x dx$$

$$dv = \lambda e^{-\lambda x} dx \Rightarrow v = -e^{-\lambda x}$$

$$E(X) = [-x^2 e^{-\lambda x}]_{x=0}^{\infty} + 2 \int_0^{\infty} x e^{-\lambda x} dx = \frac{2}{\lambda} \int_0^{\infty} \lambda x e^{-\lambda x} dx = \frac{2}{\lambda} \left(\frac{1}{\lambda} \right) = \frac{2}{\lambda^2}$$

Ex.11: Let X be a r.v. with p.d.f. $f(x, \theta) = \frac{\theta^x e^{-\theta}}{x!}$ $x=0,1,2,\dots$ $\theta > 0$ (which is called the Poisson distribution), find $E(X)$.

Solution :

$$\begin{aligned} E(X) &= \sum_{x=0}^{\infty} x f(x) = \sum_{x=0}^{\infty} x \frac{\theta^x e^{-\theta}}{x!} \\ &= 0 * \frac{\theta^0 e^{-\theta}}{0!} + 1 * \frac{\theta^1 e^{-\theta}}{1!} + 2 * \frac{\theta^2 e^{-\theta}}{2!} + 3 * \frac{\theta^3 e^{-\theta}}{3!} + \dots \\ &= \theta e^{-\theta} \left(\frac{\theta^0}{1!} + \frac{\theta^1}{2!} + \frac{\theta^2}{3!} + \dots \right) = \theta e^{-\theta} (e^{\theta}) = \theta \end{aligned}$$

Ex.12: Let X be a r.v. with p.d.f. given by

$$f(x) = \frac{1}{x^2}, \quad x \geq 1, \text{ Does } E(X) \text{ exist?}$$

Solution : $E(X) = \int_1^{\infty} x f(x) dx = \int_1^{\infty} \frac{1}{x} dx = [\ln x]_1^{\infty} = \text{does not exist}$

Remarks:

1) If X is a continuous r.v. having the c.d.f. $F(x)$ then

$$a. E(X) = \int_0^{\infty} (1 - F(x)) dx + \int_{-\infty}^0 F(x) dx$$

$$b. E(X) = \int_0^{\infty} [1 - F(x) + F(-x)] dx$$

Proof:

$$a. E(X) = \int_{-\infty}^{\infty} x f(x) dx = \int_{-\infty}^{\infty} x F'(x) dx$$

By part integration method

$$u = x \Rightarrow du = dx$$

$$dv = F'(x) dx \Rightarrow v = F(x)$$

Hence

$$\begin{aligned} E(X) &= [x F(x)]_{-\infty}^{\infty} - \int_{-\infty}^{\infty} F(x) dx = \lim_{x \rightarrow \infty} [x F(x)] - \lim_{x \rightarrow -\infty} [x F(x)] - \int_{-\infty}^{\infty} F(x) dx \\ &= \lim_{x \rightarrow \infty} x \cdot \lim_{x \rightarrow \infty} F(x) - \lim_{x \rightarrow -\infty} x \cdot \lim_{x \rightarrow -\infty} F(x) - \int_{-\infty}^{\infty} F(x) dx \\ &= \int_0^{\infty} dx - \left[\int_{-\infty}^0 F(x) dx + \int_0^{\infty} F(x) dx \right] = \int_0^{\infty} (1 - F(x)) dx - \int_{-\infty}^0 F(x) dx \end{aligned}$$

b. (H.W.)

Ex.13: Let X be a r.v. with c.d.f. given by

$$F_X(x) = \begin{cases} 0 & , x < 0 \\ x/2 & , 0 \leq x < 2 \\ 1 & , 2 \leq x \end{cases}$$

, find $E(X)$.

Solution:

1) The first technique by depending on the definition since X is continuous r.v.

hence $f(x) = F'(x) = \frac{1}{2}$, $0 < x < 2$

Now $E(X) = \int_0^2 x f(x) dx = \frac{1}{2} \int_0^2 x dx = \frac{1}{4} [x^2]_0^2 = 1$

$$E(X) = \int_{-\infty}^0 F(x) dx + \int_0^{\infty} (1 - F(x)) dx = \int_0^2 (1 - x/2) dx = \left[(x - x^2/4) \right]_0^2 = 1$$

2) The second technique by depending on the c.d.f. (the above remark 1/a)

Now $E(X) = \int_{-\infty}^0 F(x) dx + \int_0^{\infty} (1 - F(x)) dx = \int_0^2 (1 - x/2) dx = \left[(x - x^2/4) \right]_0^2 = 1$

Variance: The variance of the function $g(X)$ of the r.v. X ,denoted by $\text{Var}(g(X))$ or $\sigma_{g(X)}^2$ is defined by

$$\text{Var}(g(X)) = E(g(X) - \mu_{g(X)})^2 = E[(g(X))^2] - [E(g(X))]^2 .$$

Properties of Expected Value

1. $\text{Var}(c) = 0$ for a constant c .
2. $\text{Var}(c g(X)) = c^2 \text{Var}(g(X))$ for a constant c .
3. $\text{Var}(c g(X) \mp b) = c^2 \text{Var}(g(X))$ where b, c are constants.

Proof:

1. $\text{Var}(c) = E(c^2) - (E(c))^2 = c^2 - c^2 = 0$

3.

$$\begin{aligned} \text{Var}(c g(X) \mp b) &= E[(c g(X) \mp b)^2] - [E(c g(X) \mp b)]^2 \\ &= E[c^2 (g(X))^2 \mp 2 b c g(X) + b^2] - [c E(g(X)) \mp b]^2 \\ &= c^2 E[(g(X))^2] \mp 2 b c E(g(X)) + b^2 \\ &\quad - c^2 [E(g(X))]^2 \pm 2 b c E(g(X)) - b^2 \\ &= c^2 \{E[(g(X))^2] - [E(g(X))]^2\} = c^2 \text{Var}(g(X)) \end{aligned}$$

Remarks:

1. If $E(X) = \mu_X \rightarrow \infty$, then we say that the variance of a r.v. X does not exist.
2. If X is a continuous r.v. having the c.d.f. $F(x)$ then

$$\text{Var}(X) = \int_0^{\infty} 2x[1 - F(x) + F(-x)] dx - \mu_X^2 \quad (\text{Prove that}).$$

Let X denote the number of...
 variance, and moment generating function of X .

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Ex. 14: Let $f(x) = 6x(1-x)$ $0 < x < 1$ be the p.d.f. of a r.v. X , find the variance of $Y = 1 - 2X$?

Solution:

Firstly we find $E(X)$ and $E(X^2)$

$$E(X) = \int_0^1 x f(x) dx = 6 \int_0^1 x(1-x) dx = 6 \left[\frac{x^2}{2} - \frac{x^3}{3} \right]_0^1 = 0.5$$

$$E(X^2) = \int_0^1 x^2 f(x) dx = 6 \int_0^1 x^2(1-x) dx = 6 \left[\frac{x^3}{3} - \frac{x^4}{4} \right]_0^1 = 0.3$$

$$\text{Var}(X) = E(X^2) - (E(X))^2 = 0.3 - 0.25 = 0.05$$

$$\text{Var}(1 - 2X) = 4 \text{Var}(X) = (4)(0.05) = 0.2$$

Exercise: Let X be a r.v. with c.p.d. $F(x) = 1 - e^{-\lambda x}$ $x, \lambda > 0$, find $E(X)$

Moment Generating Function: The expected value of $e^{t g(X)}$ is defined to be the moment generating function (m.g.f.) of $g(X)$ of the r.v. X if the expected value exists for every value of t some interval $-h < t < h$; $h > 0$. The moment generating function of $g(X)$, denoted by $M_{g(X)}(t)$, is

$$M_{g(X)}(t) = E(e^{t g(X)}) = \begin{cases} \sum_{v \in X} e^{t g(v)} f(v) & \text{if } X \text{ a discrete r.v.} \\ \int_{-\infty}^{\infty} e^{t g(x)} f(x) dx & \text{if } X \text{ a continuous r.v.} \end{cases}$$

where $f(x)$ is the p.d.f. of X .

Properties of Moment Generating Function:

1) $E(X) = M'_X(0)$

2) $\text{Var}(X) = M''_X(0) - (M'_X(0))^2$

Proof:

1. $M'_X(t) = \frac{dM_X(t)}{dt} = E(X e^{tX}) \Rightarrow M'_X(0) = E(X)$

2. $M''_X(t) = \frac{d^2 M_X(t)}{dt^2} = E(X^2 e^{tX}) \Rightarrow M''_X(0) = E(X^2)$

$\text{Var}(X) = E(X^2) - (E(X))^2 = M''_X(0) - (M'_X(0))^2$

$f(x) = \frac{dF(x)}{dx} = -\lambda e^{-\lambda x}$

$E(X) = \int_0^{\infty} x \cdot f(x) dx$

$= \int_0^{\infty} x \cdot (-\lambda e^{-\lambda x}) dx$

=

Ex.15: Let X be a r.v. with p.d.f. given by $f(x, \lambda) = \lambda e^{-\lambda x}$ $\lambda, x > 0$, find the moment generating function (m.g.f.) of X .

Solution:

$$M_X(t) = E(e^{tx}) = \int_0^{\infty} e^{tx} f(x) dx = \int_0^{\infty} e^{tx} \lambda e^{-\lambda x} dx = \lambda \int_0^{\infty} e^{-(\lambda-t)x} dx$$

$$= \frac{-\lambda}{\lambda-t} [e^{-(\lambda-t)x}]_0^{\infty} = \frac{-\lambda}{\lambda-t} (0-1)$$

$$= \frac{\lambda}{\lambda-t}, (\lambda-t > 0 \Rightarrow t < \lambda)$$

Ex.16: Let X be a r.v. with p.d.f. $f(x) = f(x, n, p) = \begin{cases} \binom{n}{x} p^x q^{n-x} & x = 0, 1, \dots, n \\ 0 & \text{o.w.} \end{cases}$ where the

two parameters n, p satisfy $n = 1, 2, \dots, 0 \leq p \leq 1$ and $q = 1-p$. (which is called the Bernoulli distribution), find $M_X(t)$, $E(X)$ and $\text{Var}(X)$

Solution:

$$M_X(t) = E(e^{tx}) = \sum_{x=0}^n e^{tx} f(x) = \sum_{x=0}^n \binom{n}{x} e^{tx} p^x q^{n-x}$$

$$= \binom{n}{0} e^{0t} p^0 q^n + \binom{n}{1} e^{1t} p^1 q^{n-1} + \binom{n}{2} e^{2t} p^2 q^{n-2} + \dots + \binom{n}{n} e^{nt} p^n q^0$$

$$= \binom{n}{0} q^n + \binom{n}{1} (e^t p)^1 q^{n-1} + \binom{n}{2} (e^t p)^2 q^{n-2} + \dots + \binom{n}{n} (e^t p)^n$$

$$= (q + pe^t)^n$$

Now $M'_X(t) = n p e^t (pe^t + q)^{n-1} \Rightarrow E(X) = M'_X(0) = n p (p + q)^{n-1} = n p$ (since $q = 1-p$)

$$M''_X(t) = n(n-1)(pe^t)^2 (pe^t + q)^{n-2} + n p e^t (pe^t + q)^{n-1}$$

$$\Rightarrow E(X^2) = M''_X(0) = n(n-1)p^2 + n p$$

hence

$$\text{Var}(X) = E(X^2) - (E(X))^2 = n(n-1)p^2 + n p - n^2 p^2 = np(1-p) = npq$$

Characteristic Function: The expected value of $e^{itg(X)}$ is defined to be the characteristic function of $g(X)$ of the r.v. X if the expected value exists for every value of t some interval $-h < t < h; h > 0, i = \sqrt{-1}$. The characteristic function of $g(X)$, denoted by $\phi_{g(X)}(t)$, is

$$\phi_{g(X)}(t) = E(e^{itg(X)}) = \begin{cases} \sum_{x \in \mathcal{S}_X} e^{itg(x)} f(x) & , \text{ if } X \text{ a discrete r.v.} \\ \int_{-\infty}^{\infty} e^{itg(x)} f(x) dx & , \text{ if } X \text{ a continuous r.v.} \end{cases}$$

where $f(x)$ is the p.d.f. of X .

Properties of Characteristic Function:

- 1) $\phi_{g(X)}(t) = E(e^{itg(X)}) = E(\cos(gtX) + i\sin(gtX))$
- 2) $-1 \leq \phi_{g(X)}(t) \leq 1 \quad \forall -h < t < h, h > 0$
- 3) $\phi'_X(0) = iE(X) = iM'_X(0)$, $\phi''_X(0) = i^2 E(X^2) = i^2 M''_X(0)$
- 4) $\phi''_X(0) - (\phi'_X(0))^2 = i^2 \text{Var}(X)$

Proof:

$$2. \phi'_X(t) = \frac{d\phi_X(t)}{dt} = E(iX e^{itX}) \Rightarrow \phi'_X(0) = iE(X) = iM'_X(0)$$

$$\phi''_X(t) = \frac{d^2\phi_X(t)}{dt^2} = E(i^2 X^2 e^{itX}) \Rightarrow \phi''_X(0) = i^2 E(X^2) = i^2 M''_X(0)$$

✓ **Ex.17:** Let X be a r.v. with p.d.f. $f(x) = f(x, n, p) = \binom{10}{x} (1/3)^x (2/3)^{10-x}$, $x = 0, 1, \dots, 10$, find

$\phi_X(t)$, $E(2X-1)$ and $\text{Var}(2-X/5)$.

Solution:

$$\begin{aligned} \phi_X(t) &= E(e^{itX}) = \sum_{x=0}^{10} e^{itx} f(x) = \sum_{x=0}^{10} \binom{10}{x} e^{itx} (1/3)^x (2/3)^{10-x} \\ &= \binom{10}{0} e^{0it} (1/3)^0 (2/3)^{10} + \binom{10}{1} e^{it} (1/3)^1 (2/3)^9 + \dots + \binom{10}{10} e^{i10t} (1/3)^{10} (2/3)^0 \\ &= \binom{10}{0} (2/3)^{10} + \binom{10}{1} (e^{it} (1/3))^1 (2/3)^9 + \dots + \binom{10}{10} (e^{it} (1/3))^{10} \\ &= \left(\frac{2}{3} + \frac{1}{3} e^{it} \right)^{10} \end{aligned}$$

Now

$$\phi'_X(t) = \frac{10}{3} i e^{it} \left(\frac{2}{3} + \frac{1}{3} e^{it} \right)^9 \Rightarrow iE(X) = \phi'_X(0) = \frac{10}{3} i \Rightarrow E(X) = \frac{10}{3}$$

Hence

$$E(2X-1) = 2E(X) - 1 = \frac{20}{3} - 1 = \frac{17}{3}$$

Also

$$\begin{aligned} \phi''_X(t) &= \frac{90}{9} i^2 e^{2it} \left(\frac{2}{3} + \frac{1}{3} e^{it} \right)^9 + \frac{10}{3} i^2 e^{it} \left(\frac{2}{3} + \frac{1}{3} e^{it} \right)^9 \\ \Rightarrow i^2 E(X^2) = \phi''_X(0) &= 10i^2 + \frac{10}{3} i^2 \Rightarrow E(X^2) = \frac{40}{3} \end{aligned}$$

Hence

$$\text{Var}(X) = E(X^2) - (E(X))^2 = \frac{40}{3} - \left(\frac{10}{3} \right)^2 = \frac{20}{3} \quad \text{and} \quad \text{Var}(2-X/5) = \frac{1}{25} \text{Var}(X) = \left(\frac{1}{25} \right) \left(\frac{20}{3} \right) = \frac{4}{15}$$

Exercises(1)

1) Let $F(x) = \begin{cases} 0 & , x < 0 \\ x/8 & , 0 \leq x < 2 \\ x^2/16 & , 2 \leq x < 4 \\ 1 & , 4 \leq x \end{cases}$ be the c.d.f. of X . Find

$\Pr(|X - \mu_x| < 2\sigma_x)$ and $\Pr(X^2 \geq 3X - 2)$ by using p.d.f. and c.d.f..

2) Let $f(x) = 2^{-x}$, $x = 1, 2, \dots$ be the p.d.f. of a r.v. X . Find :
 a) Moment generating function for $Y = a + bX$, where a and b are constants.
 b) Characteristic function for X .
 c) Mean and variance of X .

3) Let $G_X(t) = \ln(M_X(t))$ where $M_X(t)$ be moment generating function of a r.v. X . Find the mean and variance of X .

4) Let the random variable X have mean μ , standard deviation σ , and m.g.f. $M_X(t)$, $-h < t < h$. Show that

$$E\left(\frac{X - \mu}{\sigma}\right) = 0, E\left(\left(\frac{X - \mu}{\sigma}\right)^2\right) = 1, \text{ and } M_{\left(\frac{X - \mu}{\sigma}\right)}(t) = e^{-\mu t / \sigma} M_X(t), -\sigma h < t < \sigma h$$

5) Let X be a random variable of the continuous type with p.d.f. $f(x)$, which is positive provided $0 < x < b < \infty$. Show that $E(X) = \int_0^b (1 - F(x)) dx$, where $F(x)$ is the c.d.f. of X .

7) Let X be a random variable of the discrete type with p.d.f. $f(x)$, which is positive on the nonnegative integers. Show that $E(X) = \sum_{x=1}^{\infty} (1 - F(x))$, where $F(x)$ is the c.d.f. of X .

7) Let X have the p.d.f. $f(x) = 1/N$, $x = 1, 2, 3, \dots, N$. Show that the m.g.f. is

$$M_X(t) = \begin{cases} \frac{e^t(1 - e^{Nt})}{N(1 - e^t)} & t \neq 0 \\ 1 & t = 0 \end{cases}$$

8) Find the moments of the distribution that has m.g.f. $M_X(t) = (1 - t)^{-3}$, $t < 1$. (Hint: Find the MacLaurin's series for $M_X(t)$).

9) Let X be a random variable have p.d.f $f(x) = 1/b$, $0 \leq x \leq b$ such that $\Pr(X < M_X''(0)) = 1/12$, compute the value b .

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10) Let X denote the numbers obtained when one dice is throw. Compute the mean, variance, and moment generating function of X .

11) Let X have p.d.f. $f(x) = \frac{1}{\lambda} e^{-x/\lambda}$, $x > 0$, $\lambda > 0$. Find m.g.f, mean, and variance of X .

12) If X is a random variable has $f(x) = \frac{\lambda^x e^{-\lambda}}{x!}$, $x = 0, 1, 2, 3, \dots$, $\lambda > 0$ such that $\Pr(X = 1) = \Pr(X = 2)$, compute $\Pr(X \geq 1)$.

13) Let X have a probability density function $f(x)$ that is positive at $x = -1, 0, 1$ and is zero elsewhere. If $f(0) = 1/2$, compute $E(X^2)$.

14) Given $f(x) = \frac{4}{x} f(x-1)$, $x = 1, 2, \dots$. Find the probability density function $f(x)$.