

MATHEMATICAL STATISTICS

CHAPTER

Lecture Notes

1

Random Variables & Probability Distributions

A **random variable** is a function defined on Sample Space which assigns each element of \mathcal{S} to a real number.

Example Let, a fair coin is tossed three times and let X be the number of Heads obtained. Then;

$$S = \{HHH, HHT, HTH, THH, HTT, THT, TTH, TTT\}$$

$$X(HHH) = 3$$

$$X(HHT) = X(HTH) = X(THH) = 2$$

$$X(HTT) = X(THT) = X(TTH) = 1$$

$$X(TTT) = 0$$

The **probability mass function (pmf)** of a discrete random variable $f(x)$ assigns probabilities to the values X can take. We call a random variable X **discrete** if it takes discrete values (usually Integers) on a countable set. We have; **Numerical Value**

Random Variable $P(X=x) = f(x)$ and if $f(x)$ is a valid pmf,

(i) $f(x) \geq 0 \rightarrow$ **Nonnegative Probabilities**

(ii) $\sum_x f(x) = 1 \rightarrow$ **Total Probability is 1.**

1.48. For each of the following, find the constant c so that $f(x)$ satisfies the condition of being a p.d.f. of one random variable X .

(a) $f(x) = c\left(\frac{2}{3}\right)^x$, $x = 1, 2, 3, \dots$, zero elsewhere.

(b) $f(x) = cx$, $x = 1, 2, 3, 4, 5, 6$, zero elsewhere.

1.49. Let $f(x) = x/15$, $x = 1, 2, 3, 4, 5$, zero elsewhere, be the p.d.f. of X . Find $\Pr(X = 1 \text{ or } 2)$, $\Pr\left(\frac{1}{2} < X < \frac{5}{2}\right)$, and $\Pr(1 \leq X \leq 2)$.

1.48) b) $f(x) = cx$ $x = 1, 2, 3, 4, 5, 6$

(i) $f(x) \geq 0$ for $c \geq 0$

(ii) $\sum_{x=1}^6 f(x) = \sum_{x=1}^6 cx = c \cdot \sum_{x=1}^6 x = c \cdot \frac{6 \cdot 7}{2} = 21c = 1$
 $c = \frac{1}{21}$

a) $f(x) = c \cdot \left(\frac{2}{3}\right)^x$ $x = 1, 2, 3, \dots$

$f(x) \geq 0$ for $c \geq 0$

$\sum_{x=1}^{\infty} f(x) = \sum_{x=1}^{\infty} c \cdot \left(\frac{2}{3}\right)^x = c \cdot \sum_{x=1}^{\infty} \left(\frac{2}{3}\right)^x = c \cdot \left(\frac{1}{1 - \frac{2}{3}}\right) = 3c = 1$
 $c = \frac{1}{3}$

Handwritten note: $\sum_{r=0}^{\infty} x^r = \frac{1}{1-x}$

1.49) $f(x) = \begin{cases} \frac{x}{15} & x = 1, 2, 3, 4, 5 \\ 0 & \text{otherwise} \end{cases}$

$P(X=1 \text{ OR } 2) = P(X=1) + P(X=2) = f(1) + f(2) = \frac{1}{15} + \frac{2}{15} = \frac{1}{5}$

$P\left(\frac{1}{2} < X < \frac{5}{2}\right) = P(0.5 < X < 2.5) = P(X=1) + P(X=2) = \frac{1}{5}$

$P(1 \leq X \leq 2) = P(X=1) + P(X=2) = \frac{1}{5}$

* Note that: $P(X \leq 3) = f(1) + f(2) + f(3)$

$P(X < 3) = f(1) + f(2)$

Cumulative Distribution Function (cdf)

The cumulative distribution function $F(x)$ is;

$$F(x) = P(X \leq x)$$

For a discrete random variable X ,

$$F(b) = \sum_{-\infty}^b f(x)$$

Given $F(x)$, it can be a cdf if it satisfies the following.

(i) $\lim_{x \rightarrow -\infty} F(x) = 0$ and $\lim_{x \rightarrow \infty} F(x) = 1$

(ii) $\lim_{h \rightarrow 0^+} F(x+h) = F(x) \rightarrow F(x)$ is right continuous

(iii) $a \leq b \Rightarrow F(a) \leq F(b) \rightarrow F(x)$ is NONdecreasing

Note that for a and b are integers and X is a random variable that take integer values;

* $P(a < X \leq b) = P(X \leq b) - P(X \leq a) = F(b) - F(a)$

* $P(X=a) = f(a) = F(a) - F(a^-)$

1.50. Let $f(x)$ be the p.d.f. of a random variable X . Find the distribution function $F(x)$ of X and sketch its graph along with that of $f(x)$ if:

(a) $f(x) = 1, x = 0$, zero elsewhere.

(b) $f(x) = \frac{1}{3}, x = -1, 0, 1$, zero elsewhere.

(c) $f(x) = x/15, x = 1, 2, 3, 4, 5$, zero elsewhere.

1.51. Let us select five cards at random and without replacement from an ordinary deck of playing cards.

(a) Find the p.d.f. of X , the number of hearts in the five cards.

(b) Determine $\Pr(X \leq 1)$.

1.52. Let X equal the number of heads in four independent flips of a coin. Using certain assumptions, determine the p.d.f. of X and compute the probability that X is equal to an odd number.



1.50) a)
$$F(x) = \begin{cases} 0 & x < 0 \\ 1 & x \geq 0 \end{cases}$$

b)

x	-1	0	1
$f(x)$	$\frac{1}{3}$	$\frac{1}{3}$	$\frac{1}{3}$
$F(x)$	$\frac{1}{3}$	$\frac{2}{3}$	1

\downarrow \downarrow \downarrow
+ + +

$\frac{1}{3}$ $\frac{2}{3}$ $\frac{3}{3}$

$$F(x) = \begin{cases} 0 & x < -1 \\ \frac{1}{3} & -1 \leq x < 0 \\ \frac{2}{3} & 0 \leq x < 1 \\ 1 & x \geq 1 \end{cases}$$

c) For $x = 1, 2, 3, 4, 5$:

$$F(b) = P(X \leq b) = \sum_{x=1}^b f(x) = \sum_{x=1}^b \frac{1}{15} = \frac{1}{15} \cdot \sum_{x=1}^b 1 = \frac{b \cdot (b+1)}{30}$$

Then:

$$F(x) = \begin{cases} 0 & x < 1 \\ \frac{11x \cdot (11x+1)}{30} & 1 \leq x < 5 \\ 1 & x \geq 5 \end{cases}$$

1.51)

a) 13 Hearts
39 Nonhearts
52 Cards

Select $n=5$ cards
 X : Number of hearts

$$f(x) = \frac{\binom{13}{x} \binom{39}{5-x}}{\binom{52}{5}}$$

for $x = 0, 1, 2, 3, 4, 5$

b) $P(X \leq 1) = f(0) + f(1) = \frac{\binom{13}{0} \binom{39}{5}}{\binom{52}{5}} + \frac{\binom{13}{1} \binom{39}{4}}{\binom{52}{5}} = 0,633$ ($= 0$ otherwise)

1.52) If X is a fair coin, $f(x) = \binom{4}{x} \cdot \left(\frac{1}{2}\right)^x \cdot \left(\frac{1}{2}\right)^{4-x} = \binom{4}{x} \cdot \left(\frac{1}{2}\right)^4$

$P(X \text{ is odd}) = f(1) + f(3) = \binom{4}{1} \cdot \left(\frac{1}{2}\right)^4 + \binom{4}{3} \cdot \left(\frac{1}{2}\right)^4 = 0,5$

1.53. Let X have the p.d.f. $f(x) = x/5050$, $x = 1, 2, 3, \dots, 100$, zero elsewhere.

(a) Compute $\Pr(X \leq 50)$.

(b) Show that the distribution function of X is $F(x) = [x]([x] + 1)/10100$, for $1 \leq x \leq 100$, where $[x]$ is the greatest integer in x .

1.54. Let a bowl contain 10 chips of the same size and shape. One and only one of these chips is red. Continue to draw chips from the bowl, one at a time and at random and without replacement, until the red chip is drawn.

(a) Find the p.d.f. of X , the number of trials needed to draw the red chip.

(b) Compute $\Pr(X \leq 4)$.

1.55. Cast a die a number of independent times until a six appears on the up side of the die.

(a) Find the p.d.f. $f(x)$ of X , the number of casts needed to obtain that first six.

(b) Show that $\sum_{x=1}^{\infty} f(x) = 1$.

(c) Determine $\Pr(X = 1, 3, 5, 7, \dots)$.

(d) Find the distribution function $F(x) = \Pr(X \leq x)$.

$$1.53) \ a) \ P(X \leq 50) = \sum_{k=1}^{50} f(k) = \sum_{k=1}^{50} \frac{k}{5050} = \frac{1}{5050} \cdot \sum_{k=1}^{50} k = \frac{1}{5050} \cdot \frac{50 \cdot 51}{2} = 0,252$$

b) For $x = 1, 2, \dots, 100$;

$$F(x) = P(X \leq x) = \sum_{w=1}^x f(w) = \sum_{w=1}^x \frac{w}{5050} = \frac{1}{5050} \cdot \sum_{w=1}^x w = \frac{1}{5050} \cdot \frac{x(x+1)}{2}$$

$$\text{Then; } F(x) = \begin{cases} 0 & x < 1 \\ \frac{[x] \cdot [x+1]}{10100} & 0 \leq x \leq 100 \\ 1 & x > 100 \end{cases}$$

$$1.54) \ a) \ f(x) = P(X=x) = P(\text{First } x-1 \text{ trials Not red, } x^{\text{th}} \text{ is red}) \\ = \frac{9}{10} \cdot \frac{8}{9} \cdot \dots \cdot \frac{9-x+2}{10-x+2} \cdot \frac{1}{10-x} = \frac{P(9, x-1)}{P(10, x)} \quad \text{Permutation}$$

$$P(X \leq 4) = \sum_{x=1}^4 f(x) = \frac{1}{10} + \frac{9}{10} \cdot \frac{1}{9} + \frac{9}{10} \cdot \frac{8}{9} \cdot \frac{1}{8} + \frac{9}{10} \cdot \frac{8}{9} \cdot \frac{7}{8} \cdot \frac{1}{6} = 4 \cdot \frac{1}{10} = 0,4$$

A random variable is **continuous** if it takes values in a real interval. The **probability density function (pdf)** of a continuous random variable $f(x)$ is used to find probability of X to be between two numbers. $f(x)$ is called a density function because it does NOT assign probabilities to X . Namely,

$$P(X=x) = 0$$

$$P(a < X < b) = \int_a^b f(x) dx$$

The area under the curve $f(x)$ between a and b is the probability of X between a and b . $f(x)$ to be a valid pdf, it should satisfy

- (i) $f(x) \geq 0 \rightarrow$ Nonnegative Probabilities
- (ii) $\int_{-\infty}^{\infty} f(x) dx = 1 \rightarrow$ Total Probability is 1.

Example Find c if $f(x) = \begin{cases} cx^3 & 0 < x < 1 \\ 0 & \text{o.w} \end{cases}$ is a pdf.

Answer (i) $f(x) \geq 0$ for $c \geq 0$

$$(ii) \int_{-\infty}^{\infty} f(x) dx = \int_{-\infty}^{\infty} cx^3 dx = \int_0^1 cx^3 dx = c \cdot \int_0^1 x^3 dx$$

$$= c \cdot \left[\frac{x^4}{4} \right]_0^1 = c \cdot \left(\frac{1}{4} - \frac{0}{4} \right) = \frac{c}{4} = 1$$

$$c = 4$$

$$f(x) = \begin{cases} 4x^3 & 0 < x < 1 \\ 0 & \text{o.w} \end{cases}$$

1.58. Let the probability set function $P(A)$ of the random variable X be $P(A) = \int_A f(x) dx$, where $f(x) = 2x/9$, $x \in \mathcal{A} = \{x: 0 < x < 3\}$. Let $A_1 = \{x: 0 < x < 1\}$, $A_2 = \{x: 2 < x < 3\}$. Compute $P(A_1) = \Pr[X \in A_1]$, $P(A_2) = \Pr(X \in A_2)$, and $P(A_1 \cup A_2) = \Pr(X \in A_1 \cup A_2)$.

1.58)
$$f(x) = \begin{cases} \frac{2x}{9} & 0 < x < 3 \\ 0 & \text{o.w.} \end{cases}$$

$$P(A_1) = P(0 < X < 1) = \int_0^1 f(x) dx = \int_0^1 \frac{2x}{9} dx = \frac{2}{9} \int_0^1 x dx = \frac{2}{9}$$

$$P(A_2) = P(2 < X < 3) = \int_2^3 \frac{2x}{9} dx = \frac{2}{9} \left[\frac{x^2}{2} \right]_2^3 = \frac{1}{9} (3^2 - 2^2) = \frac{5}{9}$$

$$P(A_1 \cup A_2) = P(A_1) + P(A_2) = \frac{2}{9} + \frac{5}{9} = \frac{7}{9}$$

1.64. For each of the following probability density functions of X , compute $\Pr(|X| < 1)$ and $\Pr(X^2 < 9)$.

(a) $f(x) = x^2/18$, $-3 < x < 3$, zero elsewhere.

(b) $f(x) = (x+2)/18$, $-2 < x < 4$, zero elsewhere.

1.64) a)
$$f(x) = \begin{cases} \frac{x^2}{18} & -3 < x < 3 \\ 0 & \text{o.w.} \end{cases}$$

$$P(|X| < 1) = P(-1 < X < 1) = \int_{-1}^1 \frac{x^2}{18} dx = \frac{1}{18} \left[\frac{x^3}{3} \right]_{-1}^1 = \frac{1}{27}$$

$$P(X^2 < 9) = P(-3 < X < 3) = \int_{-3}^3 f(x) dx = 1$$

b)
$$f(x) = \begin{cases} \frac{x+2}{18} & -2 < x < 4 \\ 0 & \text{o.w.} \end{cases}$$

$$P(|X| < 1) = \int_{-1}^1 \frac{x+2}{18} dx = \frac{1}{18} \left[\frac{x^2}{2} + 2x \right]_{-1}^1 = \frac{1}{18} \left(\frac{1}{2} + 2 - \frac{1}{2} + 2 \right) = \frac{2}{9}$$

$$P(X^2 < 9) = \int_{-3}^3 f(x) dx = \int_{-2}^3 f(x) dx = \frac{1}{18} \left[\frac{x^2}{2} + 2x \right]_{-2}^3 = \frac{1}{18} \left[\left(\frac{3^2}{2} + 6 \right) - \left(\frac{2^2}{2} - 4 \right) \right] = \frac{7}{12}$$

Cumulative Distribution Function

We replace Σ by \int to find cdf of a continuous random variable X .

$$F(x) = \int_{-\infty}^x f(\omega) d\omega$$

1.69. Find the distribution function $F(x)$ associated with each of the following probability density functions. Sketch the graphs of $f(x)$ and $F(x)$.

- (a) $f(x) = 3(1-x)^2$, $0 < x < 1$, zero elsewhere.
 (b) $f(x) = 1/x^2$, $1 < x < \infty$, zero elsewhere.
 (c) $f(x) = \frac{1}{3}$, $0 < x < 1$ or $2 < x < 4$, zero elsewhere.

1.69) a) $f(x) = \begin{cases} 3 \cdot (1-x)^2 & 0 < x < 1 \\ 0 & \text{o.w.} \end{cases}$

For $0 < x < 1$; $F(x) = \int_0^x f(\omega) d\omega = \int_0^x 3 \cdot (1-\omega)^2 d\omega = -3 \int_1^{1-x} t^2 dt$

$1-\omega = t$
 $-d\omega = -dt$

$$= 3 \int_{1-x}^1 t^2 dt = 3 \cdot \left. \frac{t^3}{3} \right|_{1-x}^1 = 1 - (1-x)^3$$

$$F(x) = \begin{cases} 0 & x \leq 0 \\ 1 - (1-x)^3 & 0 < x < 1 \\ 1 & x \geq 1 \end{cases}$$

b) For $x > 1$, $F(x) = \int_1^x f(\omega) d\omega = \int_1^x \frac{1}{\omega^2} d\omega = \left. -\frac{1}{\omega} \right|_1^x = -\frac{1}{x} + 1$

$$F(x) = \begin{cases} 0 & x < 1 \\ 1 - \frac{1}{x} & x \geq 1 \end{cases}$$

$$c) f(x) = \begin{cases} \frac{1}{3} & 0 < x < 1 \text{ OR } 2 < x < 4 \end{cases}$$

For $0 < x < 1$; $F(x) = \int_0^x \frac{1}{3} dw = \frac{x}{3}$ and $F(1) = \frac{1}{3}$

For $2 < x < 4$; $F(x) = \frac{1}{3} + \int_2^x \frac{1}{3} dw = \frac{1}{3} + \frac{1}{3} w \Big|_2^x = \frac{1}{3} + \left(\frac{x}{3} - \frac{2}{3}\right) = \frac{x}{3} - \frac{1}{3}$

$$F(x) = \begin{cases} 0 & x < 0 \\ \frac{x}{3} & 0 \leq x < 1 \\ \frac{1}{3} & 1 \leq x < 2 \\ \frac{x}{3} - \frac{1}{3} & 2 \leq x < 4 \\ 1 & x \geq 4 \end{cases}$$

1.71. Given the distribution function

$$\begin{aligned} F(x) &= 0, & x < -1, \\ &= \frac{x+2}{4}, & -1 \leq x < 1, \\ &= 1, & 1 \leq x. \end{aligned}$$

Sketch the graph of $F(x)$ and then compute: (a) $\Pr(-\frac{1}{2} < X \leq \frac{1}{2})$; (b) $\Pr(X=0)$; (c) $\Pr(X=1)$; (d) $\Pr(2 < X \leq 3)$.

1.71) a) $\Pr(-\frac{1}{2} < X \leq \frac{1}{2}) = F(\frac{1}{2}) - F(-\frac{1}{2}) = \frac{\frac{1}{2}+2}{4} - \frac{-\frac{1}{2}+2}{4} = 0.625$

b) $\Pr(X=0) = 0$ c) $\Pr(X=1) = 0$ (since X is continuous)

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

Example 19 Show that $F(x) = \begin{cases} 0 & x < 1 \\ 1 - x^{-k} & x \geq 1 \end{cases}$ is a valid cdf for $k > 0$

Answer 19 (i) $\lim_{x \rightarrow -\infty} F(x) = \lim_{x \rightarrow -\infty} 0 = 0$

$$\lim_{x \rightarrow \infty} F(x) = \lim_{x \rightarrow \infty} (1 - x^{-k}) = 1 - \underbrace{\lim_{x \rightarrow \infty} \frac{1}{x^k}}_{=0} = 1$$

(ii) $\lim_{h \rightarrow 0^+} F(x+h) = \lim_{h \rightarrow 0^+} 1 - (x+h)^{-k} = 1 - x^{-k} = F(x)$

($F(x)$ is right continuous for $x=1$ also)

(iii) $F(a) \leq F(b) \Rightarrow 1 - a^{-k} \leq 1 - b^{-k}$
 $b^{-k} \leq a^{-k}$
 $b^k \geq a^k \Rightarrow a \leq b$

$F(x)$ is NON decreasing.

Then, $F(x)$ is a valid cdf.

b) Find $P(X < 7)$ and $P(X > 10)$

Answer 19 $P(X < 7) = P(X \leq 7) = F(7) = 1 - 7^{-k}$
 $P(X > 10) = 1 - P(X \leq 10) = 1 - F(10) = 10^{-k}$

c) find $f(x)$

Answer 19 $f(x) = \frac{d}{dx} F(x) = \frac{d}{dx} (1 - x^{-k}) = k \cdot x^{-(k+1)}$



pdf of a random variable $Y=g(X)$

given pdf of a random variable $X: f_x(x)$, we can find pdf of $Y=g(X): f_y(y)$ using the following technique, which is called **distribution-function technique**.

(i) Find $F_x(x) = \int_{-\infty}^x f_x(w) dw$

(ii) Find $F_y(y) = P(Y \leq y) = P(g(X) \leq y)$
 $= P(X \leq g^{-1}(y)) = F_x(g^{-1}(y))$

(iii) Find ~~the~~ $f_y(y) = \frac{d}{dy} F_x(g^{-1}(y))$

1.73. Let $f(x) = x/6$, $x = 1, 2, 3$, zero elsewhere, be the p.d.f. of X . Find the distribution function and the p.d.f. of $Y = X^2$.

Hint: Note that X is a random variable of the discrete type.

1.74. Let $f(x) = (4-x)/16$, $-2 < x < 2$, zero elsewhere, be the p.d.f. of X .

- (a) Sketch the distribution function and the p.d.f. of X on the same set of axes.
 (b) If $Y = |X|$, compute $\Pr(Y \leq 1)$.
 (c) If $Z = X^2$, compute $\Pr(Z \leq \frac{1}{4})$.

1.75. Let X have the p.d.f. $f(x) = 2x$, $0 < x < 1$, zero elsewhere. Find the distribution function and p.d.f. of $Y = X^2$.

1.76. Let X have the p.d.f. $f(x) = 4x^3$, $0 < x < 1$, zero elsewhere. Find the distribution function and p.d.f. of $Y = -2 \ln X^4$.

1.73)

X	1	2	3
$f(x)$	$\frac{1}{6}$	$\frac{2}{6}$	$\frac{3}{6}$
$Y = X^2$	1	4	9
$f_y(y)$	$\frac{1}{6}$	$\frac{2}{6}$	$\frac{3}{6}$



1.74) a) $f(x) = \begin{cases} \frac{4-x}{16} & -2 < x < 2 \\ 0 & \text{o.w.} \end{cases}$

For $-2 < x < 2$:

$$F(x) = \int_{-2}^x \frac{4-w}{16} dw = -\frac{1}{16} \int_0^{4-x} t dt = \frac{1}{16} \cdot \frac{t^2}{2} \Big|_0^{4-x} = \frac{36 - (4-x)^2}{32}$$

$4-w \Rightarrow t$
 $-dw \Rightarrow dt$

$$F(x) = \begin{cases} 0 & x \leq -2 \\ \frac{36 - (4-x)^2}{32} & -2 < x < 2 \\ 1 & x \geq 2 \end{cases}$$

b) $P(Y \leq 1) = P(|X| \leq 1) = P(-1 \leq X \leq 1) = F(1) - F(-1)$

$$= \frac{36-9}{32} - \frac{36-25}{32} = \frac{1}{2}$$

$$P(Z \leq \frac{1}{4}) = P(X^2 \leq \frac{1}{4}) = P(-\frac{1}{2} < X < \frac{1}{2}) = F(\frac{1}{2}) - F(-\frac{1}{2}) = \frac{8}{32} = \frac{1}{4}$$

1.75) $f(x) = \begin{cases} 2x & 0 < x < 1 \\ 0 & \text{o.w.} \end{cases} \Rightarrow F(x) = \begin{cases} 0 & x \leq 0 \\ \frac{x^2}{2} & 0 < x < 1 \\ 1 & x \geq 1 \end{cases}$

$$F_Y(y) = P(Y \leq y) = P(X^2 \leq y) = P(-\sqrt{y} \leq X \leq \sqrt{y})$$

$$= F(\sqrt{y}) - F(-\sqrt{y}) = F(\sqrt{y}) = \frac{y}{2} \quad : 0 < y < 1$$

$$f_Y(y) = \frac{d}{dy} \left(\frac{y}{2} \right) = \frac{1}{2}$$

$$f_Y(y) = \begin{cases} \frac{1}{2} & 0 < y < 1 \\ 0 & \text{o.w.} \end{cases}$$

$$1.76) \quad f(x) = \begin{cases} 4x^3 & 0 < x < 1 \\ 0 & \text{o.w} \end{cases} \Rightarrow F(x) = \begin{cases} 0 & x \leq 0 \\ x^4 & 0 < x < 1 \\ 1 & x \geq 1 \end{cases}$$

$$y = -2 \ln x^4$$

$$F_y(y) = P(Y \leq y) = P(-2 \ln x^4 \leq y) = P(\ln x^4 \geq -0.5y)$$

$$= P(x^4 \geq e^{-0.5y}) = P(x \geq e^{-0.125y}) = 1 - P(x \leq e^{-0.125y})$$

$$= 1 - F(e^{-0.125y}) = 1 - (e^{-0.125y})^4 = 1 - e^{-y/2} \quad y \geq 0$$

($x=0^+ \Rightarrow y \rightarrow +\infty$ and
 $x=1 \Rightarrow y=0$)

$$f_y(y) = \frac{d}{dy} F_y(y) = \frac{d}{dy} [1 - e^{-y/2}] = \frac{1}{2} \cdot e^{-y/2}$$

Expectation of a Random Variable

Expected Value, Mean and Average have the same meaning. Expected Value of a Random Variable is its long run average.

Let, $\sum_x |x| \cdot f(x)$ converges to a limit for discrete case and $\int_{-\infty}^{\infty} |x| \cdot f(x)$ converges to a limit for continuous case. Expectation of a random variable is;

$$E(X) = \begin{cases} \sum_x x \cdot f(x) & \text{if } X \text{ is discrete} \\ \int_{-\infty}^{\infty} x \cdot f(x) dx & \text{if } X \text{ is continuous.} \end{cases}$$



Likewise, expected value of a function of random variable $X: g(x)$ is found by;

$$E[g(x)] = \begin{cases} \sum_x g(x) \cdot f(x) & \text{if } X \text{ is discrete} \\ \int_{-\infty}^{\infty} g(x) \cdot f(x) & \text{if } X \text{ is continuous.} \end{cases}$$

1.80. Let X have the p.d.f. $f(x) = (x+2)/18$, $-2 < x < 4$, zero elsewhere. Find $E(X)$, $E[(X+2)^3]$, and $E[6X - 2(X+2)^3]$.

1.80)

$$f(x) = \begin{cases} \frac{x+2}{18} & -2 < x < 4 \\ 0 & \text{o.w.} \end{cases}$$

$$E(X) = \int_{-\infty}^{\infty} x \cdot f(x) dx = \int_{-2}^4 x \cdot \frac{x+2}{18} dx = \frac{1}{18} \int_{-2}^4 (x^2 + 2x) dx = \frac{1}{18} \left(\frac{x^3}{3} - x^2 \right) \Big|_{-2}^4$$

$$= \frac{1}{18} \left[\left(\frac{4^3}{3} - 4^2 \right) - \left(-\frac{2^3}{3} - 2^2 \right) \right] = \frac{12}{18} = \frac{2}{3}$$

$$E[(X+2)^3] = \int_{-2}^4 (x+2)^3 \cdot \frac{x+2}{18} dx = \frac{1}{18} \int_{-2}^4 (x+2)^4 dx = \frac{1}{18} \int_0^6 t^4 dt$$

$x+2=t$
 $dx=dt$

$$= \frac{1}{18} \left(\frac{t^5}{5} \right) \Big|_0^6 = \frac{864}{18} = 48$$

$$E[6X - 2(X+2)^3] = 6E(X) - 2 \cdot E[(X+2)^3]$$

$$= 6 \cdot \frac{2}{3} - 2 \cdot 48 = -168$$

The last calculation is done by linearity property of Expectation operator. This follows because both Σ and \int are linear operators. For a and b are constant, we have

$$E[a \cdot g_1(X) + b \cdot g_2(X)] = a \cdot E[g_1(X)] + b \cdot E[g_2(X)]$$

Note that this is NOT true for Multiplication.

$$E[g_1(X) \cdot g_2(X)] \neq E[g_1(X)] \cdot E[g_2(X)]$$

1.84. Let X have the p.d.f. $f(x) = 3x^2$, $0 < x < 1$, zero elsewhere. Consider a random rectangle whose sides are X and $(1 - X)$. Determine the expected value of the area of the rectangle.

1.85. A bowl contains 10 chips, of which 8 are marked \$2 each and 2 are marked \$5 each. Let a person choose, at random and without replacement, 3 chips from this bowl. If the person is to receive the sum of the resulting amounts, find his expectation.

1.84) $A = X \cdot (1 - X)$

$$E(A) = E[X \cdot (1 - X)] = E[X - X^2] = E(X) - E(X^2)$$

$$E(X) = \int_0^1 x \cdot 3x^2 dx = 3 \cdot \int_0^1 x^3 dx = 3 \cdot \left. \frac{x^4}{4} \right|_0^1 = \frac{3}{4}$$

$$E(X^2) = \int_0^1 x^2 \cdot 3x^2 dx = 3 \int_0^1 x^4 dx = 3 \cdot \left. \frac{x^5}{5} \right|_0^1 = \frac{3}{5}$$

Then, $E(A) = \frac{3}{4} - \frac{3}{5} = \frac{3}{20}$

1.85)

$\$2 \rightarrow 8$
 $\$5 \rightarrow 2$
 10 chips

Select $n=3$ chips

Let $T = X_1 + X_2 + X_3$

$$P(T=24) = \frac{\binom{8}{3} \binom{2}{0}}{\binom{10}{3}} = 0,467$$

$$P(T=18) = \frac{\binom{8}{2} \binom{2}{1}}{\binom{10}{3}} = 0,467 \quad \text{and} \quad P(T=12) = \frac{\binom{8}{1} \binom{2}{2}}{\binom{10}{3}} = 0,066$$

T	24	18	12
$f_T(t)$	0,467	0,467	0,066

Then, $E(T) = \sum t \cdot f_T(t) = 24 \cdot 0,467 + 18 \cdot 0,467 + 12 \cdot 0,066 = 20,406$

1.87. Let $f(x) = 2x$, $0 < x < 1$, zero elsewhere, be the p.d.f. of X .

- (a) Compute $E(1/X)$.
- (b) Find the distribution function and the p.d.f. of $Y = 1/X$.
- (c) Compute $E(Y)$ and compare this result with the answer obtained in part (a).

Hint: Here $\mathcal{A} = \{x : 0 < x < 1\}$, find \mathcal{B} .

1.88. Two distinct integers are chosen at random and without replacement from the first six positive integers. Compute the expected value of the absolute value of the difference of these two numbers.

1.87) a) $E\left(\frac{1}{X}\right) = \int_0^1 \frac{1}{x} \cdot 2x dx = \int_0^1 2 dx = 2x \Big|_0^1 = 2$

b) $F(x) = x^2 \quad 0 < x < 1$

$F_Y(y) = P(Y \leq y) = P\left(\frac{1}{X} \leq y\right) = P\left(X \geq \frac{1}{y}\right) = 1 - P\left(X \leq \frac{1}{y}\right)$
 $= 1 - F\left(\frac{1}{y}\right) = 1 - \frac{1}{y^2}$



$$f_y(y) = \frac{d}{dy} \left[1 - \frac{1}{y^2} \right] = \frac{2}{y^3} \quad y > 1$$

$$c) E(Y) = \int_1^{\infty} y \cdot \frac{2}{y^3} dy = 2 \int_1^{\infty} \frac{1}{y^2} dy = 2 \left[-\frac{1}{y} \right]_1^{\infty} = 2(0 - (-1)) = 2$$

1.88) Table for X: Difference of two integers

I \ J	1	2	3	4	5	6
1	—	1	2	3	4	5
2	1	—	1	2	3	4
3	2	1	—	1	2	3
4	3	2	1	—	1	2
5	4	3	2	1	—	1
6	5	4	3	2	1	—

X	1	2	3	4	5
f(x)	$\frac{10}{30}$	$\frac{8}{30}$	$\frac{6}{30}$	$\frac{4}{30}$	$\frac{2}{30}$

$$\left(\text{or } f(x) = \begin{cases} \frac{6-x}{15} & x = 1, 2, 3, 4, 5 \\ 0 & \text{o.w.} \end{cases} \right)$$

$$\sum_{k=1}^n k = \frac{n \cdot (n+1)}{2}$$

$$\sum_{k=1}^n k^2 = \frac{n(n+1)(2n+1)}{6}$$

$$E(X) = \sum_x x f(x) = \sum_{x=1}^5 x \cdot \frac{6-x}{15} = \frac{1}{15} \left(6 \sum_{x=1}^5 x - \sum_{x=1}^5 x^2 \right)$$

$$= \frac{1}{15} \left(6 \cdot \frac{5 \cdot 6}{2} - \frac{5 \cdot 6 \cdot 11}{6} \right) = \frac{35}{15} = 2.33$$

Example let $f(x) = \frac{1}{\pi(1+x^2)}$ $-\infty < x < \infty$

Find $E(X)$ if it exists.

Answer Remember, $E(X)$ exists only if $\int_{-\infty}^{\infty} |x|f(x)dx$ converges to a finite limit. We have,

$$\begin{aligned} \int_{-\infty}^{\infty} |x|f(x)dx &= \int_{-\infty}^{\infty} |x| \cdot \frac{1}{\pi(1+x^2)} dx = 2 \cdot \int_0^{\infty} \frac{|x|}{\pi(1+x^2)} dx \quad (\text{since we want to integrate an even function}) \\ &= \frac{1}{\pi} \int_0^{\infty} \frac{2x}{1+x^2} dx = \frac{1}{\pi} \lim_{k \rightarrow \infty} \ln(1+x^2) \Big|_0^k = \frac{1}{\pi} \lim_{k \rightarrow \infty} \ln(k) \rightarrow \infty \end{aligned}$$

So, $E(X)$ does NOT exist.

Some Special Expectations:

$\mu = E(X)$: Mean of a random Variable X .

$\sigma^2 = \text{Var}(X)$: Variance of a random Variable X .

$\sigma^2 = E[(X-\mu)^2]$ and σ is called standard deviation.

We have;

$$\begin{aligned} \sigma^2 = \text{Var}(X) &= E[(X-\mu)^2] = E(X^2 - 2X\mu + \mu^2) \\ &= E(X^2) - 2\mu \cdot \underbrace{E(X)}_{=\mu} + \mu^2 = E(X^2) - 2\mu^2 + \mu^2 \end{aligned}$$

$$\sigma^2 = E(X^2) - \mu^2$$

1.89. Find the mean and variance, if they exist, of each of the following distributions.

(a) $f(x) = \frac{3!}{x!(3-x)!} \left(\frac{1}{2}\right)^3$, $x = 0, 1, 2, 3$, zero elsewhere.

(b) $f(x) = 6x(1-x)$, $0 < x < 1$, zero elsewhere.

(c) $f(x) = 2/x^3$, $1 < x < \infty$, zero elsewhere.

1.89) a) $f(x) = \frac{3!}{x!(3-x)!} \left(\frac{1}{2}\right)^3$ $x = 0, 1, 2, 3$

$$f(0) = \left(\frac{1}{2}\right)^3 = \frac{1}{8}; \quad f(1) = 3 \cdot \left(\frac{1}{2}\right)^3 = \frac{3}{8}; \quad f(2) = 3 \cdot \left(\frac{1}{2}\right)^3 = \frac{3}{8}; \quad f(3) = \left(\frac{1}{2}\right)^3 = \frac{1}{8}$$

$$\mu = E(X) = \sum_{x=0}^3 x \cdot f(x) = 0 \cdot \frac{1}{8} + 1 \cdot \frac{3}{8} + 2 \cdot \frac{3}{8} + 3 \cdot \frac{1}{8} = \frac{12}{8} = \frac{3}{2}$$

$$E(X^2) = \sum_{x=0}^3 x^2 \cdot f(x) = 0^2 \cdot \frac{1}{8} + 1^2 \cdot \frac{3}{8} + 2^2 \cdot \frac{3}{8} + 3^2 \cdot \frac{1}{8} = \frac{24}{8} = 3$$

$$\sigma^2 = \text{Var}(X) = E(X^2) - \mu^2 = 3 - \left(\frac{3}{2}\right)^2 = 3 - \frac{9}{4} = \frac{3}{4}$$

b) $f(x) = \begin{cases} 6x(1-x) & 0 < x < 1 \\ 0 & \text{o.w} \end{cases}$

$$\begin{aligned} \mu = E(X) &= \int_{-\infty}^{\infty} x \cdot f(x) dx = \int_0^1 x \cdot 6x(1-x) dx = 6 \int_0^1 (x^2 - x^3) dx \\ &= 6 \cdot \left[\frac{x^3}{3} - \frac{x^4}{4} \right]_0^1 = 6 \cdot \left(\frac{1}{3} - \frac{1}{4} \right) = \frac{1}{2} \end{aligned}$$

$$E(X^2) = \int_0^1 x^2 \cdot 6x(1-x) dx = 6 \int_0^1 (x^3 - x^4) dx = 6 \cdot \left(\frac{x^4}{4} - \frac{x^5}{5} \right)_0^1 = \frac{6}{20} = \frac{3}{10}$$

$$\sigma^2 = \text{Var}(X) = E(X^2) - \mu^2 = \frac{3}{10} - \left(\frac{1}{2}\right)^2 = \frac{3}{10} - \frac{1}{4} = \frac{2}{40} = \frac{1}{20}$$



c) Note that, $E(|X|) = E(X)$ since X is defined for positive values. So, we do NOT need to check if $\int_{-\infty}^{\infty} |x| f(x) dx$ converges to a finite limit.

$$\mu = E(X) = \int_{-\infty}^{\infty} x \cdot f(x) dx = \int_1^{\infty} x \cdot \frac{2}{x^3} dx = 2 \int_1^{\infty} \frac{1}{x^2} dx$$

$$= -2 \cdot \lim_{k \rightarrow \infty} \left(\frac{1}{x} \right) \Big|_1^k = -2 \cdot \lim_{k \rightarrow \infty} \left(\frac{1}{k} - 1 \right) = -2 \cdot (-1) = 2$$

$$E(X^2) = \int_1^{\infty} x^2 \cdot \frac{2}{x^3} dx = 2 \int_1^{\infty} \frac{1}{x} dx = 2 \lim_{k \rightarrow \infty} \left[\ln(x) \right]_1^k$$

$$= 2 \cdot \lim_{k \rightarrow \infty} (k - 1) \rightarrow \infty$$

Then, $\text{Var}(X) \rightarrow \infty$ (does NOT exist)

Example: let $f(x) = \binom{n}{x} \cdot p^x \cdot (1-p)^{n-x}$ $x = 0, 1, 2, \dots, n$

Find $E(X)$

Answer:

$$E(X) = \sum_{x=0}^n x \cdot \binom{n}{x} \cdot p^x \cdot (1-p)^{n-x} = \sum_{x=1}^n x \cdot \binom{n}{x} p^x (1-p)^{n-x}$$

$$= \sum_{x=1}^n n \cdot \binom{n-1}{x-1} \cdot p \cdot p^{x-1} \cdot (1-p)^{n-x}$$

$$= n \cdot p \cdot \sum_{x=1}^n \binom{n-1}{x-1} \cdot p^{x-1} \cdot (1-p)^{n-x} = n \cdot p \cdot \sum_{x=0}^{n-1} \binom{n-1}{x} p^x (1-p)^{n-1-x}$$

$$= n \cdot p \cdot 1 = n \cdot p$$

$= x \cdot \frac{n!}{x!(n-x)!} = \frac{x \cdot n \cdot (n-1)!}{x \cdot (x-1)! \cdot (n-x)!} = n \cdot \binom{n-1}{x-1}$

Moment Generating Function (mgf)

* Remember;

$$E[g(x)] = \begin{cases} \sum_x g(x) f(x) & \text{if } X \text{ is discrete} \\ \int_{-\infty}^{\infty} g(x) f(x) dx & \text{if } X \text{ is continuous.} \end{cases}$$

Now, consider the function $g(t) = e^{tX}$
 Its derivatives with respect to t are;

$$g^{(1)}(t) = \frac{d g(t)}{dt} = \frac{d e^{tX}}{dt} = X \cdot e^{tX}$$

$$g^{(2)}(t) = \frac{d^2 g(t)}{dt^2} = \frac{d}{dt} (X \cdot e^{tX}) = X^2 \cdot e^{tX}$$

and in general, $g^{(k)}(t) = X^k \cdot e^{tX}$

* The expectation $E(X^k)$ is called
 k^{th} derivative \hookrightarrow " k^{th} moment of X "

let $M^{(k)}(t) = E[g^{(k)}(t)] = E[X^k \cdot e^{tX}]$

Then, $M(t)$ is called "Moment Generating Function"

because

$$M^{(k)}(0) = E(X^k)$$

$$M(t) = E(e^{tX})$$

* Note that: $\mu = E(X) = M^{(1)}(0)$ and

$$\sigma^2 = E(X^2) - \mu^2 = M^{(2)}(0) - [M^{(1)}(0)]^2$$

Moment Generating Function has the following properties:

(i) If X has mgf $M_X(t)$ and $Y = aX + b$ then

$$M_Y(t) = e^{tb} \cdot M_X(at) \text{ because}$$

$$\begin{aligned} M_Y(t) &= E(e^{Yt}) = E(e^{(aX+b)t}) = E(e^{aXt} \cdot e^{bt}) \\ &= e^{bt} \cdot E(e^{at \cdot X}) = e^{bt} \cdot M(at) \end{aligned}$$

(ii) Let $Y = X_1 + X_2 + \dots + X_n$ and X_i are independent.

Then $M_Y(t) = M_1(t) \cdot M_2(t) \cdot \dots \cdot M_n(t)$

Noting that if X and Y are independent random variables then $E[g(X) \cdot h(Y)] = E[g(X)] \cdot E[h(Y)]$, we have;

$$\begin{aligned} M_Y(t) &= E(e^{Yt}) = E(e^{(X_1 + X_2 + \dots + X_n) \cdot t}) \\ &= E(e^{X_1 t} \cdot e^{X_2 t} \cdot \dots \cdot e^{X_n t}) = E(e^{X_1 t}) \cdot E(e^{X_2 t}) \cdot \dots \cdot E(e^{X_n t}) \\ &= M_1(t) \cdot M_2(t) \cdot \dots \cdot M_n(t) \end{aligned}$$

Example: Find mgf for the Bernoulli random variable.

Answer: $X \sim \text{Bernoulli}(p)$ \rightarrow Number of success on a single trial when prob. of success is p .

$$f(x) = p^x (1-p)^{1-x}, \quad x \in \{0, 1\}$$

$$M(t) = E(e^{tX}) = \sum_{x=0}^1 e^{tx} \cdot f(x) = e^{t \cdot 0} \cdot f(0) + e^{t \cdot 1} \cdot f(1)$$

$$M(t) = (1-p) + e^t \cdot p = p \cdot e^t + (1-p)$$

Example Find the mgf of Binomial random variable and use mgf to find its mean and variance

Answer Let X_i be independent and identically distributed iid Bernoulli trials for $i=1, 2, \dots, n$

Then, if $Y = X_1 + X_2 + \dots + X_n$

$Y \sim \text{Binomial}(n, p)$ \rightarrow Number of success on n independent trials with constant probability of success.

$$f(y) = \binom{n}{y} p^y (1-p)^{n-y} \quad y=0, 1, \dots, n$$

We have,

$$M_Y(t) = M_1(t) \cdot M_2(t) \cdot \dots \cdot M_n(t) = [M(t)]^n = [p \cdot e^t + (1-p)]^n$$

$$M_Y^{(1)}(t) = n \cdot p \cdot e^t [p \cdot e^t + (1-p)]^{n-1}$$

$$* \mu = M_Y^{(1)}(0) = E(Y) = n \cdot p \cdot [p + 1 - p]^{n-1} = \underline{n \cdot p}$$

$$M_Y^{(2)}(t) = n \cdot p \cdot e^t \cdot (n-1) \cdot p \cdot [p \cdot e^t + (1-p)]^{n-2} + n \cdot p \cdot e^t \cdot [p \cdot e^t + (1-p)]^{n-1}$$

$$M_Y^{(2)}(0) = E(Y^2) = n p^2 (n-1) + n p$$

$$* \sigma^2 = E(Y^2) - \mu^2 = n^2 p^2 - n p^2 + n p - n^2 p^2 = n p - n p^2 = n p (1-p)$$

Example Find mgf of Poisson random variable and use mgf to find its mean and variance

Answer $X \sim \text{Poisson}(\lambda) \Rightarrow f(x) = \frac{e^{-\lambda} \lambda^x}{x!} \quad x=0, 1, 2, \dots$

$$M(t) = E(e^{tX}) = \sum_{x=0}^{\infty} e^{tx} \cdot \frac{e^{-\lambda} \lambda^x}{x!} = e^{-\lambda} \sum_{x=0}^{\infty} \frac{[e^t \lambda]^x}{x!}$$

$$= e^{-\lambda} \cdot e^{\lambda \cdot e^t} = \underline{e^{\lambda(e^t - 1)}}$$

$$M^{(1)}(t) = \lambda \cdot e^t \cdot e^{-\lambda[e^t-1]} \quad ; \quad M^{(2)}(t) = \lambda^2 t \cdot e^{-\lambda[e^t-1]} + \lambda \cdot e^t \cdot e^{-\lambda[e^t-1]}$$

$$\mu = E(X) = M^{(1)}(0) = \lambda \quad ; \quad E(X^2) = M^{(2)}(0) = \lambda^2 + \lambda$$

$$\sigma^2 = \text{Var}(X) = E(X^2) - \mu^2 = (\lambda^2 + \lambda) - \lambda^2 = \lambda$$

Example let X : time to complete a certain task has an Exponential Distribution with mean 10. Also let, C : cost of completing this task is given by;

$$C = 100 + 40X + 3X^2$$

Find the mean and variance of C .

Answer $X \sim \text{Exponential}(\theta)$

$$f(x) = \frac{1}{\theta} \cdot e^{-x/\theta}$$

$$M(t) = E(e^{tx}) = \int_{-\infty}^{\infty} e^{tx} \cdot f(x) dx = \int_0^{\infty} e^{tx} \cdot \frac{1}{\theta} \cdot e^{-x/\theta} dx$$

$$= \frac{1}{\theta} \cdot \int_0^{\infty} e^{-x(\frac{1}{\theta} - t)} dx = \frac{1}{\theta} \cdot \int_0^{\infty} e^{-x(1-\theta t)/\theta} dx$$

$$M(t) = \frac{1}{\theta} \cdot \frac{\theta}{1-\theta t} \cdot \int_0^{\infty} e^{-u} du$$

$= 1$ (show this)

$$\text{let } \frac{x(1-\theta t)}{\theta} = u$$

$$\frac{(1-\theta t)}{\theta} dx = du$$

$$dx = \frac{\theta}{1-\theta t} du$$

$$\underline{\underline{M(t) = (1-\theta t)^{-1}}}$$

$$M^{(1)}(t) = -1 \cdot (-\theta) \cdot (1-\theta t)^{-2} = \theta \cdot (1-\theta t)^{-2} \quad ; \quad M^{(2)}(t) = -2 \cdot \theta \cdot (-\theta) \cdot (1-\theta t)^{-3}$$

$$\text{Likewise, } M_3^{(3)} = 3! \cdot \theta^3 \cdot (1-\theta t)^{-4} \text{ and } M^{(4)}(t) = 4! \cdot \theta^4 \cdot (1-\theta t)^{-5} = 24 \theta^4 \cdot (1-\theta t)^{-5}$$

$$\text{So; } E(X) = \theta \quad ; \quad E(X^2) = 2\theta^2 \quad ; \quad E(X^3) = 6\theta^3 \text{ and } E(X^4) = 24\theta^4$$

$$X \sim \text{Exponential} (\theta = 10)$$

$$C = 100 + 40X + 3X^2$$

$$\begin{aligned} E(C) &= E(100 + 40X + 3X^2) = 100 + 40E(X) + 3E(X^2) \\ &= 100 + 40 \cdot 10 + 3 \cdot 2 \cdot 10^2 = \underline{\underline{1100}} \end{aligned}$$

$$\begin{aligned} \text{Var}(C) &= \text{Var}(100 + 40X + 3X^2) = \text{Var}(40X + 3X^2) \\ &= E[(40X + 3X^2)^2] - E^2(40X + 3X^2) \end{aligned}$$

$$E(40X + 3X^2) = 40E(X) + 3E(X^2) = 40 \cdot 10 + 3 \cdot 2 \cdot 10^2 = 1000$$

$$E^2(40X + 3X^2) = 1000^2 = 1000000$$

$$\begin{aligned} E[(40X + 3X^2)^2] &= E[1600X^2 + 240X^3 + 9X^4] \\ &= 1600E(X^2) + 240E(X^3) + 9E(X^4) \\ &= 1600 \cdot 2 \cdot 10^2 + 240 \cdot 6 \cdot 10^3 + 9 \cdot 26 \cdot 10^4 = 3920000 \end{aligned}$$

$$\text{Then; } \text{Var}(C) = 3920000 - 1000000 = \underline{\underline{2920000}}$$

Example 4 Find Mean and Variance of Exponential Random Variable. Also find Skewness and Kurtosis.

Answer We have shown that for an Exponential Random Variable with parameter θ ; $E(X^k) = k! \cdot \theta^k$

$$\text{Then, } \mu = E(X) = \theta; E(X^2) = 2\theta^2; \sigma^2 = \text{Var}(X) = 2\theta^2 - \theta^2 = \theta^2$$

Skewness is defined as $SK = \frac{E[(X-\mu)^3]}{\sigma^3}$ and \rightarrow Measures Symmetry

Kurtosis is defined as $KT = \frac{E[(X-\mu)^4]}{\sigma^4}$ \rightarrow Measures Peakedness

$$\begin{aligned} E[(X-\mu)^3] &= E(X^3 - 3X^2\mu + 3X\mu^2 - \mu^3) \\ &= E(X^3) - 3\mu E(X^2) + 3\mu^2 E(X) - \mu^3 \\ &= 6\theta^3 - 3\theta \cdot 2\theta^2 + 3\theta^2 \cdot \theta - \theta^3 = 2\theta^3 \end{aligned}$$

$$\begin{aligned} E[(X-\mu)^4] &= E(X^4 - 4X^3\mu + 6X^2\mu^2 - 4X\mu^3 + \mu^4) \\ &= E(X^4) - 4\mu E(X^3) + 6\mu^2 E(X^2) - 4\mu^3 E(X) + \mu^4 \\ &= 24\theta^4 - 4\theta \cdot 6\theta^3 + 6\theta^2 \cdot 2\theta^2 - 4\theta^3 \cdot \theta + \theta^4 = 9\theta^4 \end{aligned}$$

Then; $SK = \frac{2\theta^3}{\theta^3} = 2$ and $KT = \frac{9\theta^4}{\theta^4} = 9$

Example Find mgf of Gamma distribution with parameters α and β .

Answer $X \sim \text{Gamma}(\alpha; \beta)$
 $f(x) = \frac{1}{\Gamma(\alpha)\beta^\alpha} \cdot x^{\alpha-1} \cdot e^{-x/\beta} \quad x \geq 0$

$$\begin{aligned} M(t) &= E(e^{tx}) = \int_0^\infty e^{tx} \cdot \frac{1}{\Gamma(\alpha)\beta^\alpha} \cdot x^{\alpha-1} \cdot e^{-x/\beta} dx \\ &= \int_0^\infty \frac{1}{\Gamma(\alpha)\beta^\alpha} \cdot x^{\alpha-1} \cdot e^{-x(\frac{1}{\beta} - t)} dx = \int_0^\infty \frac{1}{\Gamma(\alpha)\beta^\alpha} \cdot x^{\alpha-1} \cdot e^{-\frac{x}{\beta}(1-\beta t)} dx \end{aligned}$$

$$M(t) = \frac{1}{\beta^\alpha} \cdot (\beta')^\alpha \cdot \int_0^\infty \frac{1}{\Gamma(\alpha)(\beta')^\alpha} \cdot x^{\alpha-1} \cdot e^{-x/\beta'} dx \quad \text{let } \frac{1}{\frac{1-\beta t}{\beta}} = \beta'$$

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

$$M(t) = \frac{1}{\beta^\alpha} \cdot \frac{1}{\left(\frac{1-\beta t}{\beta}\right)^\alpha} = \frac{1}{\beta^\alpha} \cdot \frac{\beta^\alpha}{(1-\beta t)^\alpha} = (1-\beta t)^{-\alpha}$$



(iii) Moment Generating function is unique and completely determines the distribution of the random variable. Namely, if two random variables have the same mgf, they have (exactly) the same distribution.

Example Let mgf of a random variable X is given by $M(t) = \frac{4}{10} e^t + \frac{3}{10} e^{2t} + \frac{1}{10} e^{3t} + \frac{2}{10} e^{4t}$. Find pmf of this distribution.

Answer We have; $M(t) = E(e^{tX}) = \sum_x e^{tx} \cdot f(x)$

$$\text{then; } \frac{4}{10} \cdot e^t + \frac{3}{10} \cdot e^{2t} + \frac{1}{10} \cdot e^{3t} + \frac{2}{10} \cdot e^{4t} \\ = f(x_1) \cdot e^{x_1 t} + f(x_2) \cdot e^{x_2 t} + f(x_3) \cdot e^{x_3 t} + f(x_4) \cdot e^{x_4 t}$$

and so;

X	1	2	3	4
$f(x)$	$\frac{4}{10}$	$\frac{3}{10}$	$\frac{1}{10}$	$\frac{2}{10}$

Example Let $X_i \stackrel{i.i.d}{\sim} \text{Normal}(\mu, \sigma^2)$ is a random sample $i = 1, 2, \dots, n$. Find the distribution of sample mean $\bar{X} = \frac{\sum X_i}{n}$, if $M(t) = e^{\left(\frac{\sigma^2 t^2}{2} + \mu t\right)}$

Answer $X \sim \text{Normal}(\mu, \sigma^2)$

$$f(x) = \frac{1}{\sqrt{2\pi\sigma^2}} \cdot e^{-\frac{1}{2} \left(\frac{x-\mu}{\sigma}\right)^2}$$

Let $Y = X_1 + X_2 + \dots + X_n = \sum X_i$ and $\bar{X} = \frac{Y}{n}$



By property (ii);

$$M_Y(t) = M_{X_1}(t) \cdot M_{X_2}(t) \cdot \dots \cdot M_{X_n}(t) = [M(t)]^n = e^{\frac{n\sigma^2 t^2}{2} + n\mu t}$$

By property (i) with $a = \frac{1}{n}$ and $b = 0$;

$$M_{\bar{X}}(t) = M_Y\left(\frac{1}{n}t\right) = e^{\frac{\sigma^2 t^2}{2n} + \mu t} = e^{\frac{(\sigma^2/n) \cdot t^2}{2} + \mu t}$$

By property (iii);

$$\bar{X} \sim \text{Normal}\left(\mu; \frac{\sigma^2}{n}\right)$$

1.90. Let $f(x) = \left(\frac{1}{2}\right)^x$, $x = 1, 2, 3, \dots$, zero elsewhere, be the p.d.f. of the random variable X . Find the m.g.f., the mean, and the variance of X .

$$1.90) M(t) = E(e^{tX}) = \sum_{x=1}^{\infty} e^{tx} \cdot \left(\frac{1}{2}\right)^x = \sum_{x=1}^{\infty} \left(\frac{e^t}{2}\right)^x$$

$$= \sum_{x=0}^{\infty} \left(\frac{e^t}{2}\right)^x - 1 = \frac{1}{1 - \frac{e^t}{2}} - 1 = \frac{2}{2 - e^t} - 1 = \frac{e^t}{2 - e^t} \quad (\text{for } \frac{e^t}{2} < 1)$$

$$M'(t) = \frac{e^t(2 - e^t) + e^t \cdot e^t}{(2 - e^t)^2} = \frac{2e^t}{(2 - e^t)^2} \Rightarrow E(X) = M'(0) = 2 = \mu$$

$$M''(t) = \frac{2e^t \cdot (2 - e^t)^2 + 2(2 - e^t) \cdot 2e^{2t}}{(2 - e^t)^4} = \frac{2e^t(2 - e^t)(2 + e^t)}{(2 - e^t)^4}$$

$$\Rightarrow E(X^2) = M''(0) = \frac{2 \cdot (2-1)(2+1)}{(2-1)^4} = 6$$

$$\sigma^2 = \text{Var}(X) = E(X^2) - \mu^2 = 6 - 2^2 = 2$$