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## Module 5:

**Concept of Probability: Probability Mass Function, Probability density function.**

**Discrete Distribution: Binomial, Poisson's.**

**Continuous Distribution: Normal distribution, Exponential distribution.**

### Concept of Probability

#### Random Variable:

A real valued function defined on a sample space is called a Random Variable or a Discrete Random Variable.

A Random Variable assumes only a set of real values & the values which variable takes depends on the chance.

For Example:

- a) X takes only a set of discrete values 1,2,3,4,5,6.
- b) The values which x takes depends on the chance.

The set values 1,2,3,4,5,6 with their probabilities  $1/6$  is called the **Probability Distribution** of the variate x.

#### Continuous Random Variable:

When we deal with variates like weights and temperature then we know that these variates can take an infinite number of values in a given interval. Such type of variates are known as **Continuous Random Variable**.

**OR**

A Variable which is not discrete i.e. which can take infinite number of values in a given interval  $a \leq x \leq b$ , is called **Continuous Random Variable**

**Example:**  $\sin x$  between  $(0, \pi)$ , x is a **Continuous Random Variable**.

#### Probability Mass Function:

Suppose that  $X: S \rightarrow \mathcal{A}$  is a discrete random variable defined on a sample space  $S$ . Then the probability mass function  $p(x): \mathcal{A} \rightarrow [0, 1]$  for  $X$  is defined as:

- a)  $P(x_i) \geq 0$ , for every  $i=1,2,3..$
- b)  $\sum_{i=1}^{\infty} p(x_i) = 1$

The sum of probabilities over all possible values of a discrete random variables must be

equal to 1.

Thinking of probability as mass helps to avoid mistakes since the physical mass is conserved as is the total probability for all hypothetical outcomes  $x$ .

- The following exponentially declining distribution is an example of a distribution with an infinite number of possible outcomes—all the positive integers:

$$p(x_i) = \frac{1}{2^i}, i = 1, 2, 3, \dots$$

Despite the infinite number of possible outcomes, the total probability mass is  $1/2 + 1/4 + 1/8 + \dots = 1$ , satisfying the unit total probability requirement for a probability distribution.

### Probability Density Function:

Let  $X$  be a continuous random variable and let the probability of  $X$  falling in the infinite interval  $(x - \frac{1}{2}dx, x + \frac{1}{2}dx)$  be expressed by  $f(x)dx$ , i.e.

$$P(x - \frac{1}{2}dx, x + \frac{1}{2}dx) = f(x)dx$$

Where  $f(x)$  is a continuous function of  $X$  & satisfies the following condition:

- $f(x) \geq 0$
- $\int_a^b f(x)dx = 1$  if  $a \leq x \leq b$

Then the function is called probability density function of the continuous random variable  $X$ .

### Continuous Probability Distribution:

The Probability distribution of continuous random variate is called the continuous probability distribution and it is expressed in terms of probability density function.

### Cumulative Distribution Function:

The probability that the value of a random variate  $X$  is 'x or less than x' is called the

Cumulative distribution function of  $X$  and is usually denoted by  $F(x)$ . and it is given by

$$F(x) = P(X \leq x) = \sum_{x \leq x_i} p(x_i)$$

The cumulative distribution function of a continuous random variable is given by

$$F(x) = P(X \leq x) = \int_{-\infty}^x f(x) dx$$

Some properties of Cumulative Distribution Function:

a)  $F(-\infty) = 0$

b)  $F(X)$  is non-decreasing function

c) For a distribution variate

$$P(a < x < b) = F(b) - F(a)$$

d)  $F(+\infty) = 1$

e)  $F(x)$  is a discontinuous function for a discontinuous variate and  $F(x)$  is continuous function for a continuous variate.

### Examples:

1) Let  $X$  be a random variable with PDF given by

$$f(x) = \begin{cases} cx^2 & |x| \leq 1 \\ 0 & \text{otherwise} \end{cases}$$

a. Find the constant  $c$

b. Find  $EX$  and  $\text{Var}(X)$

c. Find  $P(X \geq 12)$ .

Solution: To find  $c$ , we can use  $\int_{-\infty}^{\infty} f(x) dx = 1$ :

$$1 = \int_{-1}^1 cx^2 dx$$

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

Therefore  $C = \frac{3}{2}$

To find  $EX$ , we can write  $\int_{-1}^1 xf(x) dx = 0$

In fact, we could have guessed  $EX=0$  because the PDF is symmetric around  $x=0$ . To find  $\text{Var}(X)$ , we have

$$\begin{aligned}\text{Var}(X) &= EX^2 - (EX)^2 = EX^2 \\ &= \int_{-1}^1 x^2 f(x) dx \\ &= 3/5\end{aligned}$$

**To Find  $P(X \geq 1/2)$ :**

$$\begin{aligned}P(X \geq 1/2) &= \frac{3}{2} \int_{1/2}^1 x^2 dx \\ &= 7/16.\end{aligned}$$

**Example:** If  $f(x) = cx^2, 0 < x < 1$ . Find the value of  $c$  and determine the probability that  $\frac{1}{3} < x < \frac{1}{2}$

Solution: By property of p.d.f. we have,  $\int_0^1 f(x) dx = 1$

So  $\int_0^1 cx^2 dx = 1$ , or  $c \left[ \frac{x^3}{3} \right]_0^1 = 1$ , so  $c = 3$

Consequently  $f(x) = 3x^2 : 0 < x < 1$

$$\text{Again } P\left(\frac{1}{3} < X < \frac{1}{2}\right) = \int_{1/3}^{1/2} 3x^2 dx = \frac{1}{8} - \frac{1}{27} = \frac{19}{216}$$

**Example:** For the distribution  $dF = \sin x dx, 0 \leq x \leq \pi/2$ . Find Mode and Median.

Solution: Here  $f(x) = \sin x, 0 \leq x \leq \frac{\pi}{2}$

(a) For Mode:  $f'(x) = 0$  &  $f''(x) < 0$ ,  $f'(x) = 0 \Rightarrow \cos x = 0 \Rightarrow x = \frac{\pi}{2}$  &

$[f''(x)]_{x=\frac{\pi}{2}} = -1 < 0$ , Hence mode =  $\frac{\pi}{2}$

Let  $M_d$  be median, then  $\int_0^{M_d} \sin x dx = \frac{1}{2} \Rightarrow M_d = \pi/3$

(b) Mean =  $\mu_1' = \int_0^{\pi/2} (x-0)f(x) dx = \int_0^{\pi/2} x \sin x dx = 1$  &

Variance =  $\mu_2 = \int_0^{\pi/2} (x-1)^2 \sin x dx = \pi - 3$

**Continuous random variable** – infinite number of values with no gaps between the values. [You might consider drawing a line, the sweeping hand on a clock, or the analog speedometer on a car.]

In this section, we restrict our discussion to discrete probability distributions. Each probability distribution must satisfy the following two conditions.

1.  $\sum P(x) = 1$  where  $x$  assumes all possible values of the random variable
2.  $0 \leq P(x) \leq 1$  for every value of  $x$

As we found the mean and standard deviation with data in descriptive statistics, we can find the mean and standard deviation for probability distributions by using the following formulas.

1.  $\mu = \sum [x \cdot P(x)]$  **mean** of probability distribution
2.  $\sigma^2 = \sum [(x - \mu)^2 \cdot P(x)]$  **variance** of probability distribution
3.  $\sigma^2 = \sum [x^2 \cdot P(x)] - \mu^2$  **variance** of probability distribution
4.  $\sigma = \sqrt{\sum [x^2 \cdot P(x)] - \mu^2}$  **standard deviation** of probability distribution

## Theoretical Distributions

**Definition :** When frequency distribution of some universe are not based on actual observation or experiments , but can be derived mathematically from certain predetermined hypothesis , then such distribution are said to be theoretical distributions.

**Types of Theoretical Distributions:** Following two types of Theoretical Distributions are usually used in statistics:

- 1) Discrete Probability Distribution
  - a) Binomial Distribution
  - b) Poisson Distribution
- 2) Continuous Probability Distribution
  - Normal Distribution

### Binomial Distribution:

1. The procedure has a **fixed number of trials**. [n trials]
2. The trials must be **independent**.
3. Each trial is in **one of two mutually exclusive categories**.
4. The **probabilities remain constant** for each trial.



### Notations:

$P(\text{success}) = P(S) = p$  probability of success in one of the n trials

$P(\text{failure}) = P(F) = 1 - p = q$  probability of failure in one of the n trials

$n$  = fixed number of trials;  $x$  = number of successes, where  $0 \leq x \leq n$

$P(x)$  = probability of getting exactly  $x$  successes among the  $n$  trials

$P(x \leq a)$  = probability of getting  $x$ -values less than or equal to the value of  $a$ .

$P(x \geq a)$  = probability of getting  $x$ -values greater than or equal to the value of  $a$ .

NOTE: Success (failure) does not necessarily mean good (bad).

**Formula for Binomial Probabilities:**  $P(x) = \frac{n!}{(n-x)!x!} p^x q^{n-x}$  for  $x=0,1,2,\dots,n$

Factorial definition:  $n! = n(n-1)(n-2)\dots 2 \cdot 1$ ;  $0! = 1$ ;  $1! = 1$

**Example (Formula):** Find the probability of 2 successes of 5 trials when the probability of success is 0.3.

$$P(x=2) = \frac{5!}{(5-2)!2!} 0.3^2 0.7^{5-2} = \frac{5 \cdot 4 \cdot 3!}{3! \cdot 2!} (0.09)(0.343) = 10(0.03087) = 0.3087$$

**Moment about the origin:**

1) **First moment about the origin:**

$$\mu'_1 = \sum_{r=0}^n r \cdot ({}^n C_r) p^r q^{n-r}$$

$$= np$$

2) **Second moment about the origin:**

$$\mu'_2 = \sum_{r=0}^n r^2 \cdot ({}^n C_r) p^r q^{n-r}$$

$$= npq + n^2 p^2$$

**Moment about the Mean:**

1) First moment about the mean is 0.

2) Second moment about the mean or variance is given by  $= npq$

$$\text{Standard deviation} = \sqrt{npq}$$

**Examples:**

(1) Six dice are thrown 729 times. How many times do you expect at least three dice to show a five or six?

**Solution:** ) We know that when a die is thrown, the probability to show a 5 or 6  $= 2/6 = 2/3 = p$  (say)

$$q = 1 - p = 1 - (1/3) = 2/3$$

The probability to show a 5 or 6 in at least 3 dice

$$= \sum_{x=3}^6 p(x) = p(3) + p(4) + p(5) + p(6), \text{ where } p(x) \text{ is the probability to show 5 or 6}$$

$$= {}^6 C_3 q^3 p^3 + {}^6 C_4 p^4 q^2 + {}^6 C_5 p^5 q + {}^6 C_6 p^6 = \frac{233}{729} = p \text{ (say)}$$

SO the required no.  $= np = 233$

(2) The mean and variance of a binomial variate are 16 & 8. Find i)  $P(X=0)$

ii)  $P(X \geq 2)$

$$\text{Mean} = np = 16$$

$$\text{Variance} = npq = 8$$

$$npq / np = 8/16 = 1/2$$

$$\text{ie, } q = 1/2$$

$$p = 1 - q = 1/2$$

$$np = 16 \quad \text{ie, } n = 32$$

$$\begin{aligned} \text{i) } P(X = 0) &= {}^nC_0 p^0 q^{n-0} \\ &= (1/2)^0 (1/2)^{32} \\ &= (1/2)^{32} \end{aligned}$$

$$\begin{aligned} \text{ii) } P(X \geq 2) &= 1 - P(X < 2) \\ &= 1 - P(X = 0, 1) \\ &= 1 - P(X = 0) - P(X = 1) \\ &= 1 - 33 (1/2)^{32} \end{aligned}$$



3) Six dice are thrown 729 times. How many times do you expect at least 3 dice to show a 5 or 6 ?

Solution : Here  $n = 6$ ,  $N = 729$

$$P(x \geq 3) = {}^6C_x p^x q^{n-x}$$

Let  $p$  be the probability of getting 5 or 6 with 1 dice

$$\text{ie, } p = 2/6 = 1/3$$

$$q = 1 - 1/3 = 2/3$$

$$\begin{aligned} P(x \geq 3) &= P(x = 2, 3, 4, 5, 6) \\ &= p(x=3) + p(x=4) + p(x=5) + p(x=6) \\ &= 0.3196 \end{aligned}$$

$$\text{number of times} = 729 * 0.3196 = 233$$

4) A basket contains 20 good oranges and 80 bad oranges. 3 oranges are drawn at random from this basket. Find the probability that out of 3 i) exactly 2 ii) at least 2 iii) at most 2 are good oranges.

Solution: Let  $p$  be the probability of getting a good orange

$$\text{ie, } p = \frac{80C_1}{100C_1}$$

$$= 0.8$$

$$q = 1 - 0.8 = 0.2$$

$$\text{i) } p(x=2) = 3C_2 (0.8)^2 (0.2)^1 = 0.384$$

$$\text{ii) } p(x \geq 2) = P(2) + p(3) = 0.896$$

$$\text{iii) } p(x \leq 2) = p(0) + p(1) + p(2) = 0.488$$

5) In a sampling a large number of parts manufactured by a machine, the mean number of defective in a sample of 20 is 2. Out of 1000 such samples how many would be expected to contain at least 3 defective parts.

$$n=20 \quad np=2$$

$$\text{ie, } p=1/10 \quad q=1-p=9/10$$

$$\begin{aligned} p(x \geq 3) &= 1 - p(x < 3) \\ &= 1 - p(x = 0, 1, 2) = 0.323 \end{aligned}$$

$$\begin{aligned} \text{Number of samples having at least 3 defective parts} &= 0.323 * 1000 \\ &= 323 \end{aligned}$$

The process of determining the most appropriate values of the parameters from the given observations and writing down the probability distribution function is known as fitting of the binomial distribution.

### Problems

1) Fit an appropriate binomial distribution and calculate the theoretical distribution

x :	0	1	2	3	4	5
f :	2	14	20	34	22	8

Here  $n = 5$ ,  $N = 100$

$$\text{Mean} = \frac{\sum x_i f_i}{\sum f_i} = 2.84$$

$$np = 2.84$$

$$p = 2.84/5 = 0.568$$

$$q = 0.432$$



$$p(r) = {}^5C_r (0.568)^r (0.432)^{5-r}, \quad r = 0, 1, 2, 3, 4, 5$$

Theoretical distributions are

r	p(r)	N* p(r)
0	0.0147	1.47 = 1
1	0.097	9.7 = 10
2	0.258	25.8 = 26
3	0.342	34.2 = 34
4	0.226	22.6 = 23
5	0.060	6 = 6
		Total = 100

### Poisson Distribution :

The **Poisson distribution** is a discrete distribution. It is often used as a model for the number of events (such as the number of telephone calls at a business, number of customers in waiting lines, number of defects in a given surface area, airplane arrivals, or the number of accidents at an intersection) in a specific time period.

**The mean is  $\lambda$ . The variance is  $\lambda$ .**

Therefore the P.D. is given by

$$P(r) = \frac{e^{-m} m^r}{r!} \text{ where } r=0,1,2,3\dots$$

m is the parameter which indicates the average number of events in the given time interval.

### Poisson distribution examples

1. The number of road construction projects that take place at any one time in a certain city follows a Poisson distribution with a mean of 3. Find the probability that exactly five road construction projects are currently taking place in this city. (0.100819)
2. The number of road construction projects that take place at any one time in a certain city follows a Poisson distribution with a mean of 7. Find the probability that more than four road construction projects are currently taking place in the city. (0.827008)
3. The number of traffic accidents that occur on a particular stretch of road during a month follows a Poisson distribution with a mean of 7.6. Find the probability that less than three accidents will occur next month on this stretch of road. (0.018757)
4. The number of traffic accidents that occur on a particular stretch of road during a month follows a Poisson distribution with a mean of 7. Find the probability of observing exactly three accidents on this stretch of road next month. (0.052129)

5. The number of traffic accidents that occur on a particular stretch of road during a month follows a Poisson distribution with a mean of 6.8. Find the probability that the next two months will both result in four accidents each occurring on this stretch of road. (0.009846)

**Examples:** In a certain factory turning razor blades, there is a small chance (1/500) for any blade to be defective. The blades are in packets of 10. Use Poisson's distribution to calculate the approximate number of packets containing no defective, one defective and two defective blades respectively in a consignment of 10,000 packets.

Solution: Here  $p = 1/500$ ,  $n = 10$ ,  $N = 10,000$  so  $m = np = 0.02$

Now  $e^{-m} = e^{-0.02} = 0.9802$

The respective frequencies containing no defective, 1 defective & 2 defective blades are given

As follows

$$Ne^{-m}, Ne^{-m} \cdot m, Ne^{-m} \cdot \frac{1}{2} m^2$$

i.e. 9802 ; 196; 2



### Normal Distribution:

The normal (or Gaussian) distribution is a continuous probability distribution that frequently occurs in nature and has many practical applications in statistics.

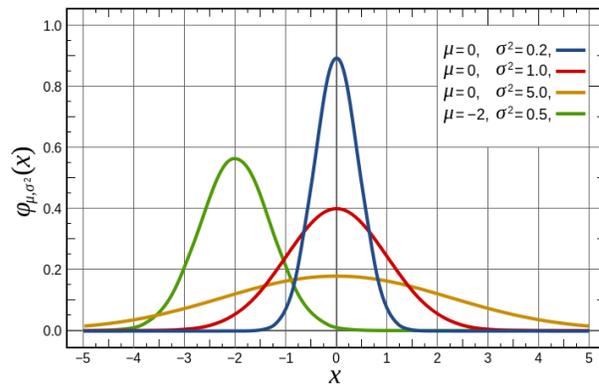
### Characteristics of a normal distribution

- Bell-shaped appearance
- Symmetrical
- Unimodal
- Mean = Median = Mode
- Described by two parameters: mean ( $\mu_x$ ) and standard deviation ( $\sigma_x$ )
- Theoretically infinite range of  $x$ : ( $-\infty < x < +\infty$ )
- The normal distribution is described by the following formula:

$$f(x; \mu, \sigma) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2} \quad -\infty < x < \infty$$

where the function  $f(x)$  defines the probability density associated with  $X = x$ . That is, the above formula is a probability density function

Because  $\mu_x$  and  $\sigma_x$  can have infinitely many values, it follows there are infinitely many normal distributions:



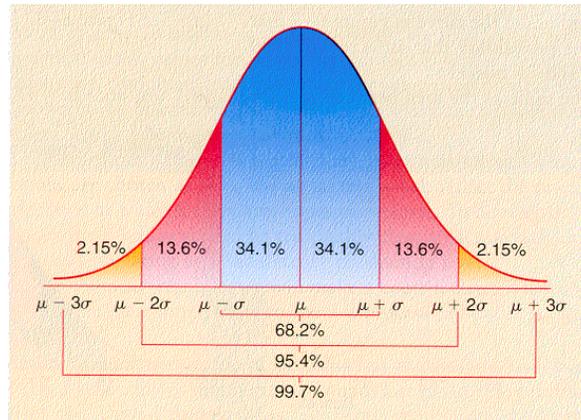
**A standard normal distribution is a normal distribution** rescaled to have  $\mu_x = 0$  and  $\sigma_x =$

1. The *pdf* is:

$$f(z;0,1) = \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}} \quad -\infty < x < \infty$$

The ordinate of the standard normal curve is no longer called  $x$ , but  $z$ .

For a normal curve, approximately 68.2%, 95.4%, and 99.7% of the observations fall within 1, 2, and 3 standard deviations of the mean, respectively.



### Areas Under the Normal Curve

By standardizing a normal distribution, we eliminate the need to consider  $\mu_x$  and  $\sigma_x$ ; we have a standard frame of reference.

### Areas Under the Standard Normal Curve

$X$  ( $x$  values) of a normal distribution map into  $Z$  ( $z$ -values) of a standard normal distribution with a 1-to-1 correspondence.

If  $X$  is a normal random variable with mean  $\mu_x$  and  $\sigma_x$ , then the standard normal variable (normal deviate) is obtained by:

$$z = \frac{x - \mu_x}{\sigma_x}$$

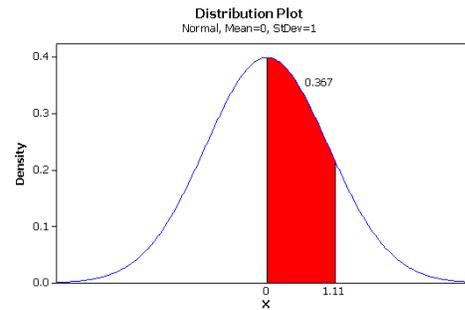
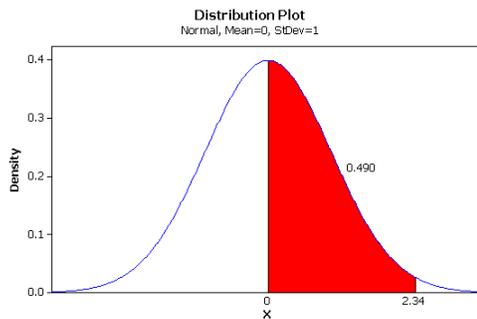
**Example 1:** What is the probability that  $Z$  falls  $z = 1.11$  and  $z = 2.34$ ?

$$\Pr(1.11 < z < 2.34)$$

$$= \text{area from } z = 2.34 \text{ to } z = 1.11$$

$$= \text{area from } (-\infty \text{ to } z = 2.34) \text{ minus area from } (-\infty \text{ to } z = 1.11)$$

$$= .9904 - .8665 = .1239$$



**Note:** figures above should also shade region from  $-\infty$  to 0.

### Table of the standard normal distribution values ( $z \leq 0$ )

$-z$	0.00	0.01	0.02	0.03	0.04	0.05	0.06	0.07	0.08	0.09
0.0	0.50000	0.49601	0.49202	0.48803	0.48405	0.48006	0.47608	0.47210	0.46812	0.46414
0.1	0.46017	0.45621	0.45224	0.44828	0.44433	0.44038	0.43644	0.43251	0.42858	0.42466
0.2	0.42074	0.41683	0.41294	0.40905	0.40517	0.40129	0.39743	0.39358	0.38974	0.38591
0.3	0.38209	0.37828	0.37448	0.37070	0.36693	0.36317	0.35942	0.35569	0.35197	0.34827
0.4	0.34458	0.34090	0.33724	0.33360	0.32997	0.32636	0.32276	0.31918	0.31561	0.31207
0.5	0.30854	0.30503	0.30153	0.29806	0.29460	0.29116	0.28774	0.28434	0.28096	0.27760
0.6	0.27425	0.27093	0.26763	0.26435	0.26109	0.25785	0.25463	0.25143	0.24825	0.24510
0.7	0.24196	0.23885	0.23576	0.23270	0.22965	0.22663	0.22363	0.22065	0.21770	0.21476
0.8	0.21186	0.20897	0.20611	0.20327	0.20045	0.19766	0.19489	0.19215	0.18943	0.18673
0.9	0.18406	0.18141	0.17879	0.17619	0.17361	0.17106	0.16853	0.16602	0.16354	0.16109
1.0	0.15866	0.15625	0.15386	0.15151	0.14917	0.14686	0.14457	0.14231	0.14007	0.13786
1.1	0.13567	0.13350	0.13136	0.12924	0.12714	0.12507	0.12302	0.12100	0.11900	0.11702
1.2	0.11507	0.11314	0.11123	0.10935	0.10749	0.10565	0.10384	0.10204	0.10027	0.09853

1.3	0.09680	0.09510	0.09342	0.09176	0.09012	0.08851	0.08692	0.08534	0.08379	0.08226
1.4	0.08076	0.07927	0.07780	0.07636	0.07493	0.07353	0.07215	0.07078	0.06944	0.06811
1.5	0.06681	0.06552	0.06426	0.06301	0.06178	0.06057	0.05938	0.05821	0.05705	0.05592
1.6	0.05480	0.05370	0.05262	0.05155	0.05050	0.04947	0.04846	0.04746	0.04648	0.04551
1.7	0.04457	0.04363	0.04272	0.04182	0.04093	0.04006	0.03920	0.03836	0.03754	0.03673
1.8	0.03593	0.03515	0.03438	0.03363	0.03288	0.03216	0.03144	0.03074	0.03005	0.02938
1.9	0.02872	0.02807	0.02743	0.02680	0.02619	0.02559	0.02500	0.02442	0.02385	0.02330
2.0	0.02275	0.02222	0.02169	0.02118	0.02068	0.02018	0.01970	0.01923	0.01876	0.01831
2.1	0.01786	0.01743	0.01700	0.01659	0.01618	0.01578	0.01539	0.01500	0.01463	0.01426
2.2	0.01390	0.01355	0.01321	0.01287	0.01255	0.01222	0.01191	0.01160	0.01130	0.01101
2.3	0.01072	0.01044	0.01017	0.00990	0.00964	0.00939	0.00914	0.00889	0.00866	0.00842
2.4	0.00820	0.00798	0.00776	0.00755	0.00734	0.00714	0.00695	0.00676	0.00657	0.00639
2.5	0.00621	0.00604	0.00587	0.00570	0.00554	0.00539	0.00523	0.00509	0.00494	0.00480
2.6	0.00466	0.00453	0.00440	0.00427	0.00415	0.00403	0.00391	0.00379	0.00368	0.00357
2.7	0.00347	0.00336	0.00326	0.00317	0.00307	0.00298	0.00289	0.00280	0.00272	0.00264
2.8	0.00256	0.00248	0.00240	0.00233	0.00226	0.00219	0.00212	0.00205	0.00199	0.00193
2.9	0.00187	0.00181	0.00175	0.00170	0.00164	0.00159	0.00154	0.00149	0.00144	0.00140
3.0	0.00135	0.00131	0.00126	0.00122	0.00118	0.00114	0.00111	0.00107	0.00104	0.00100
3.1	0.00097	0.00094	0.00090	0.00087	0.00085	0.00082	0.00079	0.00076	0.00074	0.00071
3.2	0.00069	0.00066	0.00064	0.00062	0.00060	0.00058	0.00056	0.00054	0.00052	0.00050
3.3	0.00048	0.00047	0.00045	0.00043	0.00042	0.00040	0.00039	0.00038	0.00036	0.00035
3.4	0.00034	0.00033	0.00031	0.00030	0.00029	0.00028	0.00027	0.00026	0.00025	0.00024
3.5	0.00023	0.00022	0.00022	0.00021	0.00020	0.00019	0.00019	0.00018	0.00017	0.00017

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### Table of the standard normal distribution values ( $z \geq 0$ )

<b>z</b>	<b>0.00</b>	<b>0.01</b>	<b>0.02</b>	<b>0.03</b>	<b>0.04</b>	<b>0.05</b>	<b>0.06</b>	<b>0.07</b>	<b>0.08</b>	<b>0.09</b>
<b>0.0</b>	0.50000	0.50399	0.50798	0.51197	0.51595	0.51994	0.52392	0.52790	0.53188	0.53586
<b>0.1</b>	0.53983	0.54380	0.54776	0.55172	0.55567	0.55962	0.56356	0.56749	0.57142	0.57535
<b>0.2</b>	0.57926	0.58317	0.58706	0.59095	0.59483	0.59871	0.60257	0.60642	0.61026	0.61409
<b>0.3</b>	0.61791	0.62172	0.62552	0.62930	0.63307	0.63683	0.64058	0.64431	0.64803	0.65173
<b>0.4</b>	0.65542	0.65910	0.66276	0.66640	0.67003	0.67364	0.67724	0.68082	0.68439	0.68793
<b>0.5</b>	0.69146	0.69497	0.69847	0.70194	0.70540	0.70884	0.71226	0.71566	0.71904	0.72240
<b>0.6</b>	0.72575	0.72907	0.73237	0.73565	0.73891	0.74215	0.74537	0.74857	0.75175	0.75490
<b>0.7</b>	0.75804	0.76115	0.76424	0.76730	0.77035	0.77337	0.77637	0.77935	0.78230	0.78524
<b>0.8</b>	0.78814	0.79103	0.79389	0.79673	0.79955	0.80234	0.80511	0.80785	0.81057	0.81327
<b>0.9</b>	0.81594	0.81859	0.82121	0.82381	0.82639	0.82894	0.83147	0.83398	0.83646	0.83891
<b>1.0</b>	0.84134	0.84375	0.84614	0.84849	0.85083	0.85314	0.85543	0.85769	0.85993	0.86214
<b>1.1</b>	0.86433	0.86650	0.86864	0.87076	0.87286	0.87493	0.87698	0.87900	0.88100	0.88298
<b>1.2</b>	0.88493	0.88686	0.88877	0.89065	0.89251	0.89435	0.89617	0.89796	0.89973	0.90147
<b>1.3</b>	0.90320	0.90490	0.90658	0.90824	0.90988	0.91149	0.91308	0.91466	0.91621	0.91774
<b>1.4</b>	0.91924	0.92073	0.92220	0.92364	0.92507	0.92647	0.92785	0.92922	0.93056	0.93189
<b>1.5</b>	0.93319	0.93448	0.93574	0.93699	0.93822	0.93943	0.94062	0.94179	0.94295	0.94408
<b>1.6</b>	0.94520	0.94630	0.94738	0.94845	0.94950	0.95053	0.95154	0.95254	0.95352	0.95449
<b>1.7</b>	0.95543	0.95637	0.95728	0.95818	0.95907	0.95994	0.96080	0.96164	0.96246	0.96327
<b>1.8</b>	0.96407	0.96485	0.96562	0.96638	0.96712	0.96784	0.96856	0.96926	0.96995	0.97062
<b>1.9</b>	0.97128	0.97193	0.97257	0.97320	0.97381	0.97441	0.97500	0.97558	0.97615	0.97670
<b>2.0</b>	0.97725	0.97778	0.97831	0.97882	0.97932	0.97982	0.98030	0.98077	0.98124	0.98169
<b>2.1</b>	0.98214	0.98257	0.98300	0.98341	0.98382	0.98422	0.98461	0.98500	0.98537	0.98574

2.2	0.98610	0.98645	0.98679	0.98713	0.98745	0.98778	0.98809	0.98840	0.98870	0.98899
2.3	0.98928	0.98956	0.98983	0.99010	0.99036	0.99061	0.99086	0.99111	0.99134	0.99158
2.4	0.99180	0.99202	0.99224	0.99245	0.99266	0.99286	0.99305	0.99324	0.99343	0.99361
2.5	0.99379	0.99396	0.99413	0.99430	0.99446	0.99461	0.99477	0.99492	0.99506	0.99520
2.6	0.99534	0.99547	0.99560	0.99573	0.99585	0.99598	0.99609	0.99621	0.99632	0.99643
2.7	0.99653	0.99664	0.99674	0.99683	0.99693	0.99702	0.99711	0.99720	0.99728	0.99736
2.8	0.99744	0.99752	0.99760	0.99767	0.99774	0.99781	0.99788	0.99795	0.99801	0.99807
2.9	0.99813	0.99819	0.99825	0.99831	0.99836	0.99841	0.99846	0.99851	0.99856	0.99861
3.0	0.99865	0.99869	0.99874	0.99878	0.99882	0.99886	0.99889	0.99893	0.99896	0.99900
3.1	0.99903	0.99906	0.99910	0.99913	0.99916	0.99918	0.99921	0.99924	0.99926	0.99929
3.2	0.99931	0.99934	0.99936	0.99938	0.99940	0.99942	0.99944	0.99946	0.99948	0.99950
3.3	0.99952	0.99953	0.99955	0.99957	0.99958	0.99960	0.99961	0.99962	0.99964	0.99965
3.4	0.99966	0.99968	0.99969	0.99970	0.99971	0.99972	0.99973	0.99974	0.99975	0.99976
3.5	0.99977	0.99978	0.99978	0.99979	0.99980	0.99981	0.99981	0.99982	0.99983	0.99983

### Practice Question:

Q.1 (a): From a pack of 52 cards ,6 cards are drawn at random. Find the probability of the following events:

(a) Three are red and three are black cards (b) three are king and three are queen

Q.1 (b): Out of 800 families with 4 children each, how many families would be expected to

have?

(a) 2 boys & 2 girls (b) at least one boy (c) no girl (d) at most 2 girls?

Assume equal probabilities for boys and girls.

Q.2(a): One bag contains 4 white, 6 red & 15 black balls and a second bag contains 11 white, 5 red & 9 black balls. One ball from each bag is drawn. Find the probability of the following events: (a) both balls are white (b) both balls are red (c) both balls are black (d) both balls are of the same colour.

Q.2(b): Find mean and variance of Binomial distribution .

Q.3(a): In a bolt factory, machines A, B & C manufacture respectively 25%, 35% & 40% of the total. Of their output 5%, 4% & 2% percent of defective bolts. A bolt is drawn at random from the product and is found to be defective. What are the probabilities that it was manufactured by machine A, B or C.

Q.3(b): In a normal distribution 31% of the items are under 45 and 8% are over 64. Find the mean and S.D. of the distribution.

Q.4(a): A sample of 100 dry battery cells tested to find the length of life produced the following results:

$$\mu = 12 \text{ hours}, \sigma = 3 \text{ hours}$$

Assuming the data to be normally distributed, what percentage of battery cells are expected to have life.



(i) more than 15 hours (ii) less than 6 hours (iii) between 10 & 14 hours

Q.4(b): Define the following:

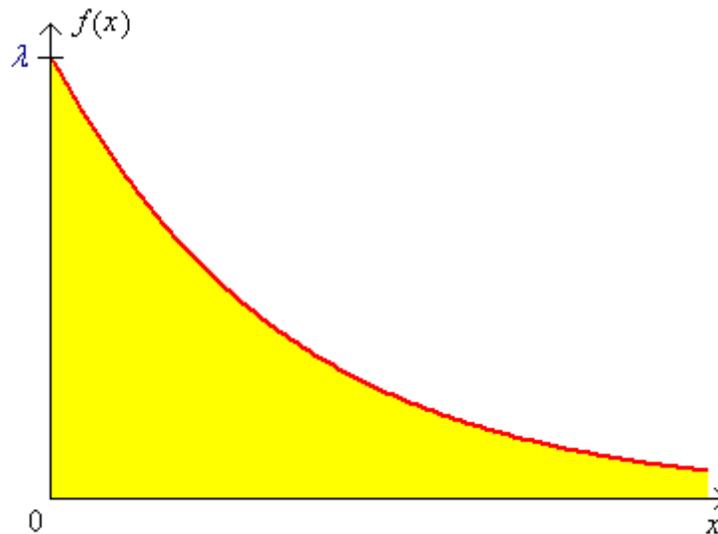
(a) Random Variable (b) Mathematical Expectation (c) Moment generating function

## The Exponential Distribution

This continuous probability distribution often arises in the consideration of lifetimes or waiting times and is a close relative of the discrete Poisson probability distribution.

The probability density function is

$$f(x) = \begin{cases} \lambda e^{-\lambda x} & (x \geq 0) \\ 0 & (x < 0) \end{cases}$$



The cumulative distribution function is

$$F(x) = P[X \leq x] = \int_{-\infty}^x f(t) dt = 0 + \left[ -e^{-\lambda t} \right]_0^x = 1 - e^{-\lambda x} \quad (x \geq 0)$$

$$\Rightarrow P[X > x] = e^{-\lambda x} \quad (x \geq 0)$$

Also  $\mu = E[X] = \frac{1}{\lambda}$  and  $\sigma = \mu$

Reason:

$$\begin{aligned}\mu &= 0 + \int_0^{\infty} x \cdot \lambda e^{-\lambda x} dx \\ &= \left[ \frac{-(\lambda x + 1)e^{-\lambda x}}{\lambda} \right]_0^{\infty} = \frac{1}{\lambda}\end{aligned}$$

$$V[X] = E[X^2] - (E[X])^2$$

$$E[X^2] = \int_0^{\infty} x^2 \cdot \lambda e^{-\lambda x} dx = \dots$$

OR

$$\begin{aligned}\sigma^2 &= \int_0^{\infty} \left(x - \frac{1}{\lambda}\right)^2 \lambda e^{-\lambda x} dx \\ &= \dots = \frac{1}{\lambda^2}\end{aligned}$$

### Example

The random quantity  $X$  follows an exponential distribution with parameter  $\lambda = 0.25$ .

Find  $\mu$ ,  $\sigma$  and  $P[X > 4]$ .

$$\mu = \sigma = \frac{1}{\lambda} = \frac{1}{.25} = 4$$

$$P[X > 4] = e^{-\lambda x} = e^{-\frac{1}{4} \times 4} = e^{-1} = .367879\dots$$

  $\approx \underline{.368}$  RGPVNOTES.IN

Note: For any exponential distribution,  $P[X > \mu] \approx .368$ .

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### Example

The waiting time  $T$  for the next customer follows an exponential distribution with a mean waiting time of five minutes. Find the probability that the next customer waits for at most ten minutes.

$$\lambda = \frac{1}{\mu} = \frac{1}{5} = .2$$

$$\begin{aligned} P[T \leq 10] &= F(10) = 1 - P[T > 10] = 1 - e^{-\frac{1}{5} \times 10} \\ &= 1 - e^{-2} = 1 - .135335... \end{aligned}$$

$$\therefore P[T \leq 10] \approx \underline{\underline{.865}}$$



Note:

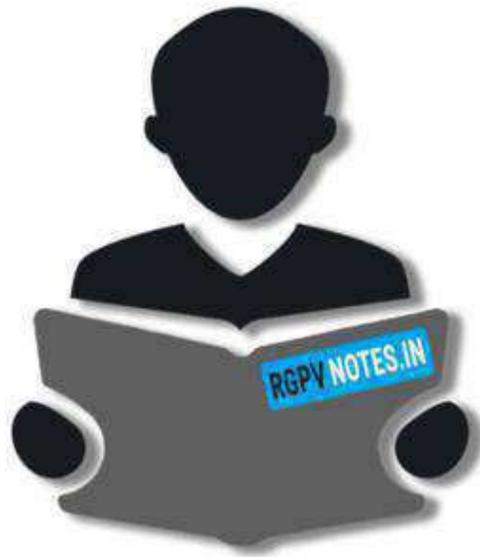
$$P[X > \mu + 2\sigma] = e^{-\lambda(\mu + 2\sigma)} = e^{-\lambda((1/\lambda) + (2/\lambda))} = e^{-3} = .049787$$

Therefore  $P[X > \mu + 2\sigma] \approx 5.0\%$  for all exponential distributions.

$$\text{Also } \mu - \sigma = \frac{1}{\lambda} - \frac{1}{\lambda} = 0 \Rightarrow P[X < \mu - \sigma] = 0 = P[X < \mu - 2\sigma]$$

Therefore  $P[|X - \mu| > 2\sigma] \approx 5.0\%$ , a result similar to the normal distribution, except that *all* of the probability is in the upper tail only.





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