## Course Title: Statistics for Economics Course Code:ECON4008 Topic: Continuous Probability Distributions M.A. Economics (2<sup>nd</sup> Semester)

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# **Theoretical Distribution**

In this topic, we will cover the following *univariate* probability distributions:

- i. Binomial Distribution
- ii. Poisson Distribution
- iii. Normal Distribution

The first two distributions are *discrete* probability distributions and the third is a *continuous* distribution.

#### Note:

- Discrete random variable: Only takes finite or countable many number of values. For example, marks obtained by students in a test, the number of defective mangoes in a basket of mangoes, number of accidents taking place on a busy road, etc.
- Continuous random variable: The random variable assume infinite and uncountable set of values. In this case, we usually talk of the value in a particular interval and not at a point. For example, the age, height or weight of students in a class are all continuous random variable.

# **Normal Distribution**

- Normal probability distribution is one of the most important continuous theoretical distributions in Statistics.
- The normal distribution was first discovered in 1733 by English mathematician De-Moivre who obtained this continuous distribution as a limiting case of the binomial distribution and applied it to problems arising in the game of chance.

#### Definition

A random variable X is said to have a normal distribution with parameters μ (called "mean") and σ (called "variance") if its density function is given by the probability law:

$$f(x; \mu, \sigma) = \frac{1}{\sigma \sqrt{2\pi}} \exp \left[ -\frac{1}{2} \left\{ \frac{x - \mu}{\sigma} \right\}^2 \right]$$
  
or 
$$f(x; \mu, \sigma) = \frac{1}{\sigma \sqrt{2\pi}} e^{-(x - \mu)^2/2\sigma^2}$$
$$-\infty < x < \infty, -\infty < \mu < \infty, \sigma > 0$$

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Where  $\pi$  and e are the constants given by:  $\pi = 22/7$ .  $\sqrt{2 \pi} = 2.5066$  and e = 2.71828.

# **Normal Distribution**

#### **Standard Normal distribution**

 If X is a random variable following normal distribution with mean μ and standard deviation σ, then the random variable Z defined as follows:

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$$Z = \frac{X - E(X)}{\sigma_x} = \frac{X - \mu}{\sigma}$$

is called the standard normal variable. We have:

$$E(Z) = E\left(\frac{X-\mu}{\sigma}\right) = \frac{1}{\sigma}E(X-\mu) = \frac{1}{\sigma}[E(X) - E(\mu)] = \frac{1}{\sigma}[\mu - \mu] = 0$$

$$Var(Z) = Var\left(\frac{X-\mu}{\sigma}\right) = \frac{1}{\sigma^2} Var(X-\mu) = \frac{1}{\sigma^2} Var(X) = \frac{1}{\sigma^2} \sigma^2 = 1$$

Therefore, the standard normal variate Z has mean 0 and standard deviation 1. Hence the probability density function (p.d.f.) of standard normal variate Z is given by:

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$$\phi(z) = \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}}, -\infty < z < \infty$$

Taking x=z,  $\mu$ =0 and  $\sigma$ =1

• The normal probability curve with mean  $\mu$  and standard deviation  $\sigma$  is given by the equation

$$f(x) = \frac{1}{\sqrt{2\pi} \cdot \sigma} e^{-(x-\mu)^2/2\sigma^2}, -\infty < x < \infty$$

> The standard normal probability curve is given by the equation:

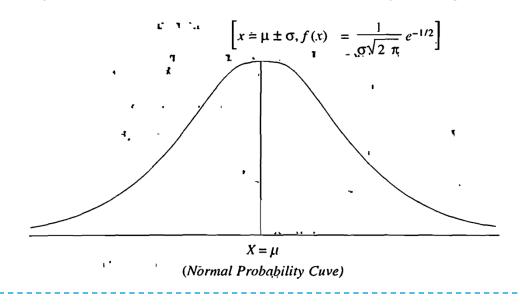
$$\phi(z) = \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}}, -\infty < z < \infty$$

- It has the following properties
- 1. The curve is bell shaped and symmetrical about the line  $x = \mu$
- 2. Mean, median and mode of the distribution coincide.
- 3. As x increases numerically, f(x) decreases rapidly, the maximum probability occurring at the point  $x = \mu$ , and given by

$$[f(x)]_{\max} = \frac{1}{\sqrt{2\pi}.\sigma}$$

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- 4. If  $\beta_1=0$  and  $\beta_2=3$
- 5.  $\mu_{2r+1} = 0$ , (r = 0, 1, 2, ...), and  $\mu_{2r} = 1.3.5$  ... (2r 1)  $\sigma^{2r}$ , (r = 0, 1, 2, ....)
- 6. Since f(x) being the probability, can never be negative. no portion of the curve lies below the x-axis.
- 7. Linear combination of independent normal variates is also a normal variate.
- 8. x-axis is an asymptote to the curve.
- 9. The points of inflexion of the curve are given by



#### 10. Area Property

 $\begin{array}{l} P \ (\mu - \sigma < X < \mu + \sigma) \ = 0.6826 \\ P \ (\mu - 2 \ \sigma < X < \mu + 2 \ \sigma) \ = 0.9544 \\ P \ (\mu - 3 \ \sigma < X < \mu + 3 \ \sigma) \ = 0.9973 \end{array}$ 

11. The following table gives the area under the normal probability curve for some important values of standard normal variate Z•

Distances from the mean ordinates in terms of $\pm \sigma$	Area under the curve
$Z = \pm 0.745$	50% = 0.50
$Z = \pm 1.00$	68-26% = 0.6826
$Z = \pm 1.96$	95% = 0·95
$Z = \pm 2.0$	95·44 % = 0·9544
$Z = \pm 2.58$	99% = 0.99
Z=±,3.0	99.73% = 0.9973

- 12. If X and Y are independent standard normal variates, then it can be easily proved that U = X+ Y and V=X -Y are Independently distributed
- 13. Area Property (Normal Probability Integral). If X~N( $\mu$ ,  $\sigma^2$ ), then the probability that random value of X will lie between  $X = \mu$  and  $X = x_1$  is given by

$$P\left(\mu < X < \dot{x}_{1}\right) = \int_{\mu}^{x_{1}} f(x) dx = \frac{1}{\sigma\sqrt{2\pi}} \int_{\mu}^{x_{1}} e^{-(x-\mu)^{2}/(2\sigma^{2})} dx$$
Put  $\frac{X-\mu}{\sigma} = Z$ , *i.e.*,  $X-\mu = \sigma Z$   
When  $X = \mu$ ,  $Z = 0$  and when  $X = x_{1}$ ,  $Z = \frac{x_{1}-\mu}{\sigma} = z_{1}$ , (say).  
 $\therefore P(\mu < X < x_{1}) = P(0 < Z < z_{1}) = \frac{1}{\sqrt{2\pi}} \int_{0}^{z_{1}} e^{-z^{2}/2} dz = \int_{0}^{z_{1}} \varphi(z) dz$   
where  $\varphi(z) = \frac{1}{\sqrt{2\pi}} e^{-z^{2}/2}$ , is the probability function of standard normal variate.

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14. The total area under normal probability curve is unity, *i.e.*,

$$\int_{-\infty}^{\infty} f(x) \, dx = \int_{-\infty}^{\infty} \varphi(z) \, dz = 1$$

#### **Reference:**

Gupta, S. C. (2015), Fundamentals of Statistics, Himalaya Publishing House.

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