

Last unit

Theory of moments, moment generating function and its properties.

Moments

Moments is used to know or identify the nature and behaviour of any data or distribution.

Types of Moments :-

- (i) Moments about mean OR Central moment (M_x)
 - (ii) Raw moments OR moments about any point OR Non-central moments (M'_x)
 - (iii) Factorial moments.
- ★ Raw moments OR Moment about any assumed mean OR about any value μ ~~is called a central moment~~

UK NON-CENTRAL MOMENTS

Definition :-

The r^{th} order raw moment about origin is denoted by M'_r and defined as

$$M'_r = E[x^r] = \begin{cases} \sum_x x^r \cdot P[x=x] & \text{Discrete} \\ \int_{-\infty}^{\infty} x^r \cdot f(x) dx & \text{Continuous} \end{cases}$$

$\therefore r = 1, 2, 3, 4$

on putting $r=1, 2, 3, 4$ we get

$$M'_1 = \text{mean} = \underline{E(x)} = \begin{cases} \sum_x x \cdot P[x=x] \\ \int_{-\infty}^{\infty} x \cdot f(x) dx \end{cases}$$

M'_2, M'_3 and M'_4

M'_1 first order

M'_1 Raw moment about origin

Central Moment :-

The γ th order moment about mean is denoted by M_γ and defined as

$$M_\gamma = E(x - \bar{x})^\gamma = E[x - E(x)]^\gamma$$

$$= \begin{cases} \sum_x (x - \bar{x})^\gamma \cdot P[x=x] ; & \text{Discrete} \\ \int_{-\infty}^{\infty} (x - \bar{x})^\gamma \cdot f(x) dx ; & \text{Continuous} \end{cases}$$

$\therefore \gamma = 1, 2, 3, 4$

on putting $\gamma = 1, 2, 3, 4$, we get

$$\begin{aligned} M_1 &= E[x - \bar{x}]' = E[x - E(x)] \\ &= E(x) - E(x) = 0 \end{aligned}$$

$$\Rightarrow M_1 = 0$$

This shows that first order central moment about mean is always zero.

$$M_2 = E(x - \bar{x})^2 = E(x - E(x))^2 [\because E(x) = \bar{x}]$$

$$\Rightarrow M_2 = \sigma^2 = \text{variance}$$

$$\boxed{M_2 = E[x - E(x)] = E(x^2) - \underline{\{E(x)\}^2}}$$

This shows that second order central moment is Variance.

$$M_3 = E[x - E(x)]^3$$

$$M_4 = E[x - E(x)]^4$$

$$\begin{aligned}
 E[x - E(x)]^2 & \stackrel{\text{Rough}}{=} E[x^2 + \{E(x)\}^2 - 2x \cdot E(x)] \\
 & = E(x^2) + \{E(x)\}^2 - 2E(x) \cdot E(x) \\
 & = E(x^2) + \{E(x)\}^2 - 2\{E(x)\}^2 \\
 & = \underline{E(x^2) - \{E(x)\}^2}
 \end{aligned}$$

Given that

$$E(x^2) = 20, \quad E(x) = 4$$

What is Variance?

$$\text{Variance} = \mu_2 = E(x^2) - \{E(x)\}^2$$

$$= 20 - (4)^2$$

$$= 20 - 16 = 4$$

Variance = 4

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

$$= \sqrt{4} = 2$$

\Rightarrow Variance = 4
Standard Deviation = 2

Moment Generating Function :-

The moment generating function of random variable X at a real parameter 't' is denoted by $M_X(t)$ and defined as

$$M_X(t) = E(e^{tx})$$

$$= \sum_x e^{tx} \cdot P[x=x] \quad \text{Discrete}$$

$$\left[\int_{-\infty}^{\infty} e^{tx} f(x) dx \right] \text{ Continuous}$$

Properties :-

(1) If $x_1, x_2, x_3, \dots, x_n$ are n independent random variables then

$$M_{x_1 + x_2 + \dots + x_n}(t) = M_{x_1}(t) \cdot M_{x_2}(t) \cdots M_{x_n}(t)$$

(2) Moment generating function is not independent of change of origin and scale

$$M_{\left(\frac{x-\mu}{\sigma}\right)}(t) = e^{-\frac{\mu t}{\sigma}} M_x(t/\sigma)$$