

Assignment

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Subject :- Probability and Statistics

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1. (i) Answer :- Random variable x is uniformly distributed in $[-2, 2]$

⇒ Probability density function

$$f(x) = \begin{cases} 1/4 & x \in [-2, 2] \\ 0 & \text{elsewhere} \end{cases}$$

Now,

$$\Rightarrow |x-1| \geq 0.5$$

$$\Rightarrow x-1 \geq 0.5 \quad \text{or} \quad x-1 \leq -0.5$$

$$\Rightarrow x \geq 1.5 \quad \text{or} \quad x \leq 0.5$$

$$\Rightarrow P(|x-1| \geq 0.5) = P(x \geq 1.5) + P(x \leq 0.5)$$

$$= \int_{1.5}^{\infty} f(x) dx + \int_{-\infty}^{0.5} f(x) dx$$

$$= \int_{1.5}^2 \frac{1}{4} dx + 0.5 \int_{-2}^{0.5} \frac{1}{4} dx$$

$$\Rightarrow \frac{0.5}{4} + \frac{2.5}{4}$$

$$\Rightarrow 3/4$$

$$P(|x-1| \geq 0.5) = 0.75 \quad \underline{\text{Ans}}$$

Q. (11)

Answer:-

The following problem is an example of Binomial Distribution with $n = 2000$, $p = 1/2$

$$\text{Let, } X \sim B(2000, 1/2)$$

$$\text{Now, } \bar{n} = np = 2000 \times \frac{1}{2} = 1000$$

$$\sigma^2 = npq = 2000 \times \frac{1}{2} \times \frac{1}{2} = 500$$

Now, according to Chebyshev's inequality

$$P(|n - \bar{n}| \geq \epsilon) \leq \frac{\sigma^2}{\epsilon^2}$$

$$\text{Let, } \epsilon = 100$$

$$P(|n - 1000| \geq 100) \leq \frac{500}{100^2}$$

$$P(X-1000 \geq 100) + P(X-1000 \leq -100) \leq \frac{1}{20}$$

$$P(X \geq 1100) + P(X \leq 900) \leq \frac{1}{20}$$

Now,

$$P(900 < X < 1100) = 1 - (P(X \geq 1100)$$

$$+ P(X \leq 900)) \geq 1 - \frac{1}{20}$$

$$P(900 < X < 1100) \geq \frac{19}{20}$$

Hence proved

Q2. (i) Answer:- Given

$$f(u) = \frac{1}{2\lambda} e^{-\frac{|u-\mu|}{\lambda}}; \quad -\infty < u < \infty, \lambda > 0$$

$$\text{mean} \quad \alpha_1 = E(X) = \int_{-\infty}^{\infty} u f(u) du = \int_{-\infty}^{\infty} \frac{u}{2\lambda} e^{-\frac{|\mu-u|}{\lambda}} du$$

$$\Rightarrow \alpha_1 = \int_{-\infty}^{\mu} \frac{u}{2\lambda} e^{\frac{\mu-u}{\lambda}} du + \int_{\mu}^{\infty} \frac{u}{2\lambda} e^{-\frac{u-\mu}{\lambda}} du$$

let,

$$I_1 = \int_{-\infty}^u \cdot \frac{u}{2d} e^{\frac{u-u}{d}} du \quad \text{and}$$

$$I_2 = \int_u^{\infty} \frac{u}{2d} e^{\frac{u-u}{d}} du$$

let,

$$\frac{u-u}{d} = z \Rightarrow u = u + dz$$

$$du = dz$$

$$\Rightarrow I_1 = \int_{-\infty}^0 \cdot \frac{1}{2d} (u + dz) e^z dz$$

$$= \frac{u}{2} \int_{-\infty}^0 e^z dz + \frac{d}{2} \cdot \int_{-\infty}^0 z e^z dz$$

$$\Rightarrow \frac{u}{2} + \frac{d}{2} \left[0 - \int_{-\infty}^0 e^z dz \right]$$

$$\Rightarrow \frac{u}{2} - \frac{d}{2}$$

$$I_1 = \frac{u}{2} - \frac{d}{2}$$

Now, let, $\frac{u-u}{d} = k \Rightarrow u = u - dk$

$$du = -dk$$

$$I_2 = \int_0^{\infty} \left(\frac{(u-\mu)u}{2\sigma^2} \right) e^{\frac{u-\mu}{\sigma}} (-1) du = \frac{\mu}{2} \int_{-\infty}^0 e^k dk$$

$$- \frac{\sigma}{2} \int_{-\infty}^0 k e^k dk$$

$$= \frac{\mu}{2} - \frac{\sigma}{2} (-1) = \frac{\mu}{2} + \frac{\sigma}{2}$$

Now, $d_1 = I_1 + I_2 = \mu$

Mean = μ Ans

Variance $\cdot u_2 = E[(x-m)^2]$ where m is mean

$$\Rightarrow u_2 = E(x^2 - 2mx + m^2)$$

$$= E(x^2) - 2mE(x) + m^2$$

$$= d_2 - 2m \cdot m + m^2$$

$$= d_2 - m^2$$

$$= d_2 - d_1^2$$

$$d_2 = E(x^2) = \int_{-\infty}^{\infty} u^2 f(u) du + \int_{-\infty}^{\infty} \frac{u^2}{2\sigma} e^{\frac{u-\mu}{\sigma}} du$$

$$+ \int_{-\infty}^{\infty} \frac{u^2}{2\sigma} e^{\frac{u-\mu}{\sigma}} du$$

$$\text{let, } I_1 = \int_{-\infty}^u \frac{u^L}{2d} e^{\frac{u-u}{d}} du$$

$$\text{and } I_2 = \int_u^{\infty} \frac{u^L}{2d} e^{\frac{u-u}{d}} du$$

$$\text{let, } \frac{u-u}{d} = z \Rightarrow u = u + dz$$

$$du = dz$$

$$\Rightarrow I_1 = \int_{-\infty}^0 \frac{(u^L + 2u dz + d^L z^2)}{2d} \times e^z \times dz$$

$$= \frac{u^L}{2} \int_{-\infty}^0 e^{-z} dz + u d \int_{-\infty}^0 z e^z dz$$

$$+ \frac{d^L}{2} \int_{-\infty}^0 z^2 e^z dz$$

$$= \frac{u^L}{2} - u d + \frac{d^L}{2} [0 - 2 \int_{-\infty}^0 z e^z dz]$$

$$I_1 = \frac{u^L}{2} - u d + d^2$$

$$\text{let, } \frac{u-u}{d} = z \Rightarrow u = u - dz$$

$$\Rightarrow du = -dz$$

$$d_2 = d_1 + \bar{L} = 2d^2 + \mu^L$$

$$\mu_2 = d_2 - d_1^L = 2d^L + \mu^L - \mu^L = 2d^L$$

$$\boxed{\sigma = \sqrt{\mu_2} = \sqrt{2}d}$$

$$= \underline{\underline{\mu_m}}$$

$$\mu_3 = E[(x-m)^3]$$

$$= E(x^3 - 3x^2m + 3m^2x - m^3)$$

$$= d_3 - 3md_2 + 3m^2m - m^3$$

$$= d_3 - 3d_1d_2 + 2d_1^3$$

$$d_3 = E(x^3) = \int_{-\infty}^{\infty} \frac{u^3}{2\pi} e^{-\frac{|u-\mu|}{\sigma}} du$$

$$= \int_{-\infty}^{\mu} \frac{u^3}{2\pi} e^{\frac{u-\mu}{\sigma}} du + \int_{\mu}^{\infty} \frac{u^3}{2\pi} e^{-\frac{u-\mu}{\sigma}} du$$

$$\text{Let, } I_1 = \int_{-\infty}^{\mu} \frac{u^3}{2\pi} e^{\frac{u-\mu}{\sigma}} du \text{ and}$$

$$I_2 = \int_{\mu}^{\infty} \frac{u^3}{2\pi} e^{-\frac{u-\mu}{\sigma}} du$$

$$\text{Let, } \frac{u-\mu}{\sigma} = z \Rightarrow u = \mu + \sigma z \quad du = \sigma dz$$

$$I_1 = \int_{-\infty}^0 \left(\frac{u^2 + 2uh + h^2}{2h} \right) e^h dh$$

$$= \frac{u^2}{2} \int_{-\infty}^0 e^h dh + u \int_{-\infty}^0 h e^h dh + \frac{1}{2} \int_{-\infty}^0 h^2 e^h dh$$

$$\Rightarrow \frac{u^2}{2} - u + \frac{1}{2} \left[0 - 2 \int_{-\infty}^0 h e^h dh \right]$$

$$\Rightarrow \frac{u^2}{2} - u + \frac{1}{2} [-2(-1)]$$

$$I_1 = \frac{u^2}{2} - u + 1$$

$$\text{let, } \frac{u-n}{r} = z \Rightarrow u = n - rz$$

$$du = -r dz$$

$$d_2 = I_1 + I_2 = 2d^2 + \mu^2$$

$$\mu_2 = d_2 - d_1^2 = 2d^2 + \mu^2 - \mu^2 = 2d^2$$

$$\boxed{\sigma = \sqrt{\mu_2} = \sqrt{2}d}$$

$$= \underline{\underline{\sigma_m}}$$

$$\mu_3 = E[(x-m)^3]$$

$$= E(x^3 - 3x^2m + 3m^2x - m^3)$$

$$= d_3 - 3md_2 + 3m^2m - m^3$$

$$= d_3 - 3d_1d_2 + 2d_1^3$$

$$d_3 = E(x^3) = \int_{-\infty}^{\infty} \frac{u^3}{2d} e^{-\frac{|u-m|}{d}} du$$

$$= \int_{-\infty}^m \frac{u^3}{2d} e^{\frac{u-m}{d}} du + \int_m^{\infty} \frac{u^3}{2d} e^{-\frac{u-m}{d}} du$$

$$\text{let, } I_1 = \int_{-\infty}^m \frac{u^3}{2d} e^{\frac{u-m}{d}} du \quad \text{and}$$

$$I_2 = \int_m^{\infty} \frac{u^3}{2d} e^{-\frac{u-m}{d}} du$$

$$\text{let, } \frac{u-m}{d} = z \quad \Rightarrow u = m + dz \quad du = d dz$$

$$I_1 = \int_{-\infty}^0 \left(\frac{u^2 + 2u + 1}{2} \right) e^u du$$

$$= \frac{u^2}{2} \int_{-\infty}^0 e^u du + u \int_{-\infty}^0 e^u du + \frac{1}{2} \int_{-\infty}^0 e^u du$$

$$\Rightarrow \frac{u^2}{2} - u + \frac{1}{2} \left[0 - 2 \int_{-\infty}^0 u e^u du \right]$$

$$\Rightarrow \frac{u^2}{2} - u + \frac{1}{2} [-2(-1)]$$

$$I_1 = \frac{u^2}{2} - u + 1$$

let, $u = n - z \Rightarrow n = u + z$

$$du = -dz$$

$$I_1 = \int_0^\infty \frac{u^3 - 3u^2 z + 3u z^2 e^z - z^3 e^z}{2z} dz$$

$$= \frac{u^3}{2} \int_{-\infty}^0 e^z dz - \frac{3u^2}{2} \int_{-\infty}^0 z e^z dz + \frac{3u}{2} \int_{-\infty}^0 z^2 e^z dz - \frac{1}{2} \int_{-\infty}^0 z^3 e^z dz$$

$$= \frac{u^3}{2} + \frac{3u^2}{2} + 3u + 1$$

$$\alpha_3 = I_1 + I_2 = u^3 + 6u^2$$

$$u_3 = \alpha_3 - 3\alpha_1 u - 2\alpha_2^2$$

$$= u^3 + 6u^2 - 3u(2u^2 + u^4) + 2u^3$$

$$= u^3 + 6u^2 - 6u^3 - 3u^5 + 2u^3$$

$$= 0$$

coefficient of skewness

$$\left[\gamma_1 = \frac{u_3}{\sigma^3} = 0 \right] \quad \frac{\mu_3}{\sigma^3}$$

2. (ii) Answer:-

$$f(x) = \frac{d}{dx} \frac{1}{(x-a)^2 + d^2} \quad -\infty < x < \infty$$

Now,

$$F(x) = \int_{-\infty}^{\infty} f(x) dx = \int_{-\infty}^x \frac{d}{dx} \frac{1}{(x-a)^2 + d^2} dx$$

$$= \frac{d}{dx} \times \frac{1}{d} \left[\tan^{-1} \left(\frac{x-a}{d} \right) \right]_{-\infty}^x$$

$$= \frac{1}{d} \left[\tan^{-1} \left(\frac{x-a}{d} \right) + \frac{\pi}{2} \right]$$

$$F\left(\frac{3}{4}\right) = \frac{3}{4}$$

$$= \frac{1}{d} \left[\tan^{-1} \left(\frac{x-a}{d} \right) + \frac{\pi}{2} \right] = \frac{3}{4}$$

$$= \tan^{-1} \left(\frac{x-a}{d} \right) = \frac{\pi}{4}$$

$$= \frac{x-a}{d} = 1$$

$$x = d + a$$

$$\frac{3}{4} = d + a$$

$$F\left(\frac{3}{4}\right) = \frac{1}{4}$$

$$\Rightarrow \frac{1}{\sigma} \left[\tan^{-1} \left(\frac{x-\mu}{\sigma} \right) + \frac{\pi}{4} \right] = \frac{1}{4}$$

$$\Rightarrow \tan^{-1} \left(\frac{x-\mu}{\sigma} \right) = -\frac{\pi}{4}$$

$$\frac{x-\mu}{\sigma} = -1$$

$$x = \mu - \sigma$$

$$Z_{1/4} = \mu - \sigma$$

Now,

$$\begin{aligned} \text{Quartile deviation} &= Z_{3/4} - Z_{1/4} \\ &= \mu + \sigma - \mu + \sigma = \sigma \frac{4}{2} \end{aligned}$$

3.7(i)

$$\text{Let, } x \sim \chi^2$$

$$f(x) = \frac{e^{-x/2} \left(\frac{x}{2} \right)^{n/2 - 1}}{2 \sqrt{\pi/2}} \quad 0 < x < \infty$$

$$= 0 \quad \text{elsewhere}$$

characteristic function $\cdot \chi(t) = \mathbb{E}(e^{itn})$

$$= \int_{-\infty}^{\infty} e^{itn} f(n) dn$$

$$= \int_0^{\infty} \frac{e^{itn} e^{-n/2} n^{n/2-1}}{2^{n/2} \sqrt{n/2}} dn$$

$$= \frac{1}{2^{n/2} \sqrt{n/2}} \int_0^{\infty} e^{itn} e^{-n/2} n^{n/2-1} dn$$

$$e^{itn} = \sum_{k=0}^{\infty} \frac{1}{k!} (itn)^k$$

$$\chi(t) = \frac{1}{2^{n/2} \sqrt{n/2}} \int_0^{\infty} \left(\sum_{k=0}^{\infty} \frac{1}{k!} (itn)^k \right) e^{-n/2} n^{n/2-1} dn$$

$$= \frac{1}{2^{n/2} \sqrt{n/2}} \left(\sum_{k=0}^{\infty} \frac{1}{k!} (it)^k \int_0^{\infty} e^{-n/2} n^{n/2+k-1} dn \right)$$

$$= \frac{1}{2^{n/2} \sqrt{n/2}} \left(\sum_{k=0}^{\infty} \frac{1}{k!} (it)^k 2^{n/2+k} \sqrt{n/2+k} \int_0^{\infty} \frac{e^{-n/2} n^{n/2+k-1}}{2^{n/2+k} \sqrt{n/2+k}} dn \right)$$

now

$\frac{e^{-u/2} u^{u+n/2-1}}{2^{u+n/2} \sqrt{u+n/2}}$ is a probability density function.

χ^2 - distribution with $2u+n$ degree of freedom

$\int_0^\infty \frac{e^{-u/2} u^{u+n/2-1}}{2^{u+n/2} \sqrt{u+n/2}} du = 1$ { Integral of probability density function over $(-\infty, \infty)$ is 1 }

$$X(t) = \frac{1}{2^{n/2} \sqrt{n/2}} \sum_{k=0}^{\infty} \frac{1}{k!} (2it)^k 2^{n/2} u^{\sqrt{n/2}+k}$$

$$= \sum_{k=0}^{\infty} \frac{1}{k!} (2it)^k 2^k \frac{\sqrt{n/2+k}}{\sqrt{n/2}}$$

$$= 1 + \sum_{k=1}^{\infty} \frac{1}{k!} (2it)^k \frac{\sqrt{n/2+k}}{\Gamma(n/2)}$$

$$= 1 + \sum_{k=1}^{\infty} \frac{1}{k!} (2it)^k \cdot \frac{\Gamma(n/2)}{\Gamma(n/2+k)}$$

$$\Rightarrow 1 + \frac{(2it)^1}{1!} \times \frac{n}{2} + \frac{(2it)^2}{2!} \times \left(\frac{n}{2}\right) \left(\frac{n}{2} + 1\right) + \dots$$

$$X(t) = (1 - 2it)^{-n/2} \underline{\underline{A_2}}$$

$$d_1 = \text{coeff of } it = n$$

$$d_2 = \text{coeff of } \frac{(it)^2}{2!} = 2^2 \left(\frac{n}{2}\right) \left(\frac{n}{2} + 1\right) = n(n+2)$$

$$\sigma^2 = d_2 - d_1^2 = n^2 + 2n - n^2 = 2n \quad \{$$

3. (ii) Ans:-

$$X \sim F(m, n)$$

$$\Rightarrow f(n) = \frac{m^{m/2} n^{n/2} \cdot n^{m/2-1}}{\beta_2\left(\frac{m}{2}, \frac{n}{2}\right) (mn+n)^{m+n}} \quad n > 0$$

$$= 0 \quad n < 0$$

$$Y = \frac{1}{X}$$

Let, $Y = 1/n$. (Monotonic on $n \in (0, \infty)$)

$$\frac{dy}{dn} = \frac{-1}{n^2} \Rightarrow \frac{dn}{dy} = -\frac{1}{y^2}$$

$$f(n) = \frac{m^{m/2} n^{n/2} n^{m/2-1}}{\beta_2\left(\frac{m}{2}, \frac{n}{2}\right) (mn+n)^{m+n/2}}$$

$$f(y) = f(n) = \frac{m^{m/2} n^{n/2} \left(\frac{1}{y}\right)^{m/2-1}}{\beta_2\left(\frac{m}{2}, \frac{n}{2}\right) (m/y+n)^{m+n/2}} \times \frac{1}{y^2}$$

$$= \frac{m^{m/2} n^{n/2} \left(\frac{1}{y}\right)^{m/2-1}}{\beta_2\left(\frac{n}{2}, \frac{m}{2}\right) (ny+m)^{m+n/2}} \times \frac{1}{y^2}$$

$$= \frac{m^{n/2} n^{n/2} \left(\frac{1}{y}\right)^{-n/2-1}}{\beta_2\left(\frac{n}{2}, \frac{m}{2}\right) (ny+m)^{\frac{n+m}{2}}} \times \frac{1}{y^2}$$

$$\frac{n^{n/2} m^{m/2} \cdot y^{n/2-1}}{\beta_2\left(\frac{n}{2}, \frac{m}{2}\right) (ny+m)^{n+m/2}} \quad \left\{ \because \beta_2\left(\frac{n}{2}, \frac{m}{2}\right) = \beta_2(n, m) \right.$$

$$\Rightarrow y \sim F(n/m) \text{ hence proved}$$

4. (i) Given $n = 20$

Sample mean $= \bar{u} = 16.3$

standard deviation of population $= \sigma = 5.2$

$$\text{Now, } P(-u_c < U < u_c) = 1 - \alpha$$

$$\begin{aligned} \text{Now, } 1 - \alpha &= 0.95 \\ &= \alpha = 0.05 \\ \frac{\alpha}{2} &= 0.025 \end{aligned}$$

$$\frac{1}{\sqrt{2\pi}} \int_0^{1.96} e^{-t^2/2} dt = 0.4750$$

$$\begin{aligned} \text{Now, } \frac{1}{\sqrt{2\pi}} \int_0^{\infty} e^{-t^2/2} dt &= \sqrt{\pi/2} \times \frac{1}{\sqrt{2\pi}} \times \sqrt{2} \\ &= \frac{1}{2} = 0.5 \end{aligned}$$

$$\therefore \frac{1}{\sqrt{2\pi}} \int_{1.96}^{\infty} e^{-t^2/2} dt = 0.025 = \frac{\alpha}{2}$$

$$\Rightarrow P(U > 1.96) = 0.025 = \frac{\alpha}{2}$$

$$\text{Now if } P(-u_e < u < u_e) = 1 - \epsilon$$

$$P(|u| < u_e) = 1 - \epsilon$$

$$P(|u| > u_e) = \epsilon$$

$$P(u > u_e) = \epsilon/2$$

Therefore the confidence interval is

$$\left(\bar{x} - \frac{6 u_e}{\sqrt{n}}, \bar{x} + \frac{6 u_e}{\sqrt{n}} \right) = (14.52, 12.18)$$

4) (ii) Ans

Now

$$s_{xy} = \frac{\text{Cov}(x, y)}{\cdot s_x \cdot s_y} \rightarrow (1)$$

$$\text{Now, } \cdot s_x^2 = E \cdot [(x - \bar{x})^2]$$

$$s_y^2 = E \cdot [(y - \bar{y})^2]$$

$$\text{Let } \cdot \text{Cov}(x, y) = E[(x - \bar{x})(y - \bar{y})]$$

$$\text{var}(x + y) = E \cdot [E(x + y) - E(x + y)]^2]$$

$$\leq E \cdot [x + y - (\bar{x} + \bar{y})]^2]$$

$$\Rightarrow E[\{(X-\bar{x}) + (Y-\bar{y})\}^2]$$

$$= E[(X-\bar{x})^2] + E[(Y-\bar{y})^2] + 2E[(X-\bar{x})(Y-\bar{y})]$$

$$= \sigma_x^2 + \sigma_y^2 + 2 \text{Cov}(x, y)$$

$$\text{var}(X+Y) = \sigma_x^2 + \sigma_y^2 + 2\rho_{xy} \sigma_x \sigma_y$$

Hence prove -

$$\text{var}(X-Y) = E[\{(X-Y) - E(X-Y)\}^2]$$

$$= E[\{X-Y - (\bar{x}-\bar{y})\}^2]$$

$$= E[\{(X-\bar{x}) - (Y-\bar{y})\}^2]$$

$$= E[(X-\bar{x})^2] + E[(Y-\bar{y})^2]$$

$$- 2E[(X-\bar{x})(Y-\bar{y})]$$

$$= \sigma_x^2 + \sigma_y^2 - 2 \text{Cov}(x, y)$$

$$\text{var}(X-Y) = \sigma_x^2 + \sigma_y^2 - 2\rho_{xy} \sigma_x \sigma_y \text{ Proved}$$

Hence prove