

ASSIGNMENT - 1

Probability & Statistics 2

Q1. $X \sim \text{Poi}(\theta)$
 $\frac{X - \theta}{\sqrt{\theta}} \sim N(0, 1)$

$$P[X_1 < \theta < X_2] = 0.95$$

$$X \sim \text{Poi}(200)$$

$$\begin{aligned} 95\% \text{CI} &= \mu \pm 1.96\sqrt{\theta} \\ &= 200 \pm 1.96\sqrt{200} \\ &= 200 + 1.96\sqrt{200} \quad \text{OR} \quad 200 - 1.96\sqrt{200} \\ &= 227.7186 \quad \text{OR} \quad 172.2814 \end{aligned}$$

$\therefore$ The interval is $(172.2814, 227.7186)$

Q2. $\bar{X} = \frac{\sum_{i=1}^n X_i}{n} = \text{sample mean}$

$$S^2 = \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X})^2 = \text{sample variance}$$

$$(i) E(\bar{X}) = E\left(\frac{\sum X_i}{n}\right) = \frac{1}{n} E(\sum X_i) = \frac{1}{n} \times n\mu = \boxed{\mu}$$

$$V(\bar{X}) = V\left(\frac{\sum X_i}{n}\right) = \frac{1}{n^2} V(\sum X_i) = \frac{n\sigma^2}{n^2} = \boxed{\frac{\sigma^2}{n}}$$

$$(ii) E(S^2) = E \left[\frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X})^2 \right]$$

$$= \frac{1}{n-1} E \left[\sum X_i^2 - n\bar{X}^2 \right]$$

$$= \frac{1}{n-1} \left[E(\sum X_i^2) - E(n\bar{X}^2) \right]$$

$$= \frac{1}{n-1} \left[E(\sigma^2 + \mu^2) - n \left(\frac{\sigma^2}{n} + \mu^2 \right) \right]$$

$$= \frac{1}{n-1} \left[n\sigma^2 + n\mu^2 - \sigma^2 - n\mu^2 \right]$$

$$= \frac{1}{(n-1)} \left[(n-1)\sigma^2 \right] = \sigma^2$$

$$(iii) \frac{(n-1)S^2}{\sigma^2} \sim \chi^2_{n-1}$$

$$\text{Var} \left[\frac{(n-1)S^2}{\sigma^2} \right] = 2(n-1)$$

$$\frac{(n-1)^2 \text{Var}[S^2]}{\sigma^4} = 2(n-1)$$

$$\boxed{\text{Var}[S^2] = \frac{2\sigma^4}{(n-1)}}$$

(iv) $n = 10$
 $\sigma^2 = 100$

$$\frac{(n-1)s^2}{\sigma^2} = 0.5$$

$$\frac{9 \times s^2}{100} = 0.5$$

$$s^2 = \frac{5}{10} \times \frac{100}{9}$$

$$s^2 = \frac{50}{9}$$

$$\therefore s = 2.357\%$$

Q 6. (ii) $P(X \geq 3 | P = 0.3)$

$$+ P(X = 0 | P = 0.3) \cdot P(Y \geq 5 | P = 0.3)$$

$$+ P(X = 1 | P = 0.3) \cdot P(Y \geq 4 | P = 0.3)$$

$$+ P(X = 2 | P = 0.3) \cdot P(Y \geq 3 | P = 0.3)$$

$$= 0.91253 \approx \boxed{91\%}$$

(iii) $P(\text{type II error})$ is the probability of accepting H_0 when H_0 is false.

$$\therefore P(\text{type II error}) = 1 - 0.91253$$

$$= 0.08747 \approx \boxed{9\% (P = 0.3)}$$

- (iv) 10 space agencies fire 120 missiles and record 40 hits

$$X \sim \text{Bin}(120, \frac{1}{3}) ; \hat{p} = \frac{40}{120} = \frac{1}{3}$$

$$\frac{\frac{x}{n} - p}{\sqrt{\frac{p(1-p)}{n}}} \sim N(0,1)$$

$$P\left(-1.2816 < \frac{\hat{p} - p}{\sqrt{\frac{p(1-p)}{n}}} < 1.2816\right)$$

$$\therefore \text{At CI } 90\% \quad p \notin (0.3333, 0.2782)$$

Q 8 (i) Unbiased estimator:

$$Eg(X) = \theta$$

$$Eg(X) = \mu \rightarrow g(E) \text{ is unbiased estimator.}$$

(ii) If $E(\theta) \neq \theta$, the estimator is biased

$$(iii) \text{MSE} = \text{variance} + \text{bias}^2$$

$$(iv) \text{MSE} \propto \frac{1}{\text{efficiency}}$$

$\therefore \hat{\theta}$ is more efficient than $\tilde{\theta}$ since the MSE of $\hat{\theta}$ is less than $\tilde{\theta}$

(v) When the MSE is more towards 0 and the n increases, then the estimator would be more efficient.

(vi) Method of moments, Maximum likelihood estimator

Q3. Let no. of claims be a random variable X .
 $X \sim \text{Bin}(4, p)$

(i) $L = \prod_{i=1}^{180} f(x_i)$

$$= [P(X=0)]^{96} \times [P(X=1)]^{75} \times [P(X=2)]^{16} \times [P(X=3)]^2 \times [P(X=4)]^1$$

$$= [{}^4C_0 \cdot p^0 (1-p)^4]^{96} \times [{}^4C_1 \cdot p^1 (1-p)^3]^{75} \times [{}^4C_2 \cdot p^2 (1-p)^2]^8$$

$$\times [{}^4C_3 \cdot p^3 (1-p)^1]^2 \times [{}^4C_4 \cdot p^4]^1$$

$$= \text{constant} \times p^{117} \times (1-p)^{603}$$

$$\log L = \log(\text{constant}) + 117 \log(p) + 603 \log(1-p)$$

$$\frac{dL}{dp} = 0 + \frac{117}{p} - \frac{603}{1-p}$$

$$\frac{dL}{dp} = \frac{117}{p} - \frac{603}{(1-p)}$$

$$\frac{dL}{dp} = 0$$

$$\therefore \frac{117}{p} = \frac{603}{1-p}$$

$$117 - 117p = 603p$$

$$720p = 117$$

$$\therefore \hat{p} = 0.1625$$

(ii) Goodness of fit test

H_0 : no. of claims conforms to the binomial model

H_1 : no. of claims does not conform to the binomial model

x	0	1	2	3	4
O_i	86	79	16	2	1
E_i	88.942	68.724	19.998	2.574	0.108

$$p = 0.1625$$

$$P(X=0) = {}^4C_0 p^0 (1-p)^4 = 0.4919$$

$$P(X=1) = {}^4C_1 p^1 (1-p)^3 = 0.3818$$

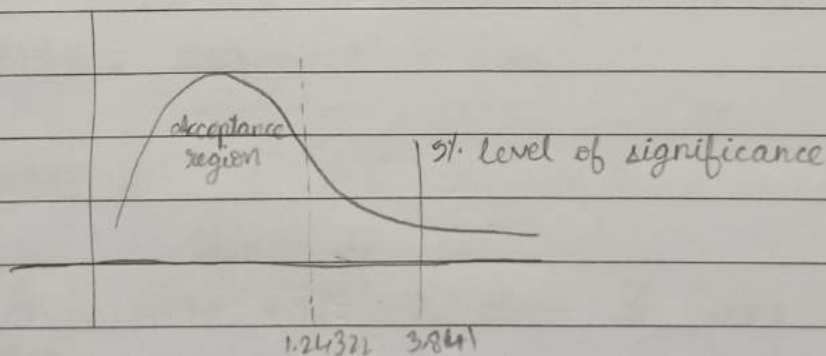
$$P(X=2) = {}^4C_2 p^2 (1-p)^2 = 0.1114$$

$$P(X=3) = {}^4C_3 p^3 (1-p)^1 = 0.0143$$

$$P(X=4) = {}^4C_4 p^4 (1-p)^0 = 0.0006$$

$$T.S. = \frac{\sum (O_i - E_i)^2}{E_i} \sim \chi^2_{5-1-1-2}$$

$$1.24322 \sim \chi^2_1$$



We have insufficient evidence to reject H_0 at 5% level of significance, therefore, no. of claims conformed to the binomial model.

Q4. $f(x) = \frac{C\beta^3}{(x+\beta)^4}$; $x > 0$, $\beta > 0$

(i) Comparing the above $f(x)$ to PDF of Pareto distribution (2 parameter version)

For pareto,

$$\text{PDF} = \frac{\alpha \lambda^\alpha}{(\lambda+x)^{\alpha+1}}$$

Thus β corresponds to λ

$$\therefore C = 3 \text{ i.e. } \alpha = 3$$

$$\therefore E(X) = \frac{\lambda}{\alpha-1} = \frac{\beta}{2}$$

$$V(X) = \frac{\alpha \lambda^2}{(\alpha-1)^2(\alpha-2)} = \frac{3\beta^2}{4}$$

(i) $E(X) = \frac{\beta}{2}$

$$E(X) = \bar{X}$$

$$\frac{\beta}{2} = \bar{X}$$

$$\hat{\beta} = 2\bar{X}$$

$$\Rightarrow \text{Bias} = E(\hat{\beta}) - \beta = E(2\bar{X}) - \beta$$

$$= 2E(\bar{X}) - \beta$$

$$= \frac{2}{n} \sum E(X_i) - \beta$$

$$= \frac{2}{n} \times n\beta - \beta = \beta$$

$$= 0 \text{ unbiased.}$$

$$\Rightarrow \text{MSE} = V(\hat{\beta}) + (\text{Bias})^2 = V(\hat{\beta}) = V(2\bar{X})$$

$$= 4V(\bar{X})$$

$$= 4V\left(\frac{\sum X_i}{n}\right)$$

$$= \frac{4}{n^2} \sum V(X_i) = \frac{4}{n^2} \times 3\beta^2 n$$

$$\therefore \text{MSE} = \frac{3\beta^2}{n}$$

As n increases, MSE tends to 0
Hence, $\hat{\beta}$ is constant.

Q5. (i) $X \sim U(a, b)$

$$E(X) = \frac{a+b}{2}$$

$$\bar{X} = \frac{a+b}{2}$$

$$\bar{X}b = 2\bar{X} - a \quad \dots (1)$$

$$V(X) = \frac{(b-a)^2}{12}$$

$$S^2 = \frac{(b-a)^2}{12}$$

$$12S^2 = (b-a)^2$$

$$2\sqrt{3}S = b-a$$

$$2\sqrt{3}S = 2\bar{X} - a - a \quad [\text{from (1)}]$$

$$2\sqrt{3}S = 2\bar{X} - 2a$$

$$\sqrt{3}S = \bar{X} - a$$

$$\hat{a} = \bar{X} - \sqrt{3}S$$

(ii) $b = 2\bar{X} - a$

$$b = 2\bar{X} - \bar{X} + \sqrt{3}S$$

$$\hat{b} = \bar{X} + \sqrt{3}S$$

(iii) sample of observations: 1, 2, 8, 50, 150

$$\bar{X} = 42.2$$

$$S = 63.57$$

$$\hat{a} = \bar{X} - \sqrt{3}S = 42.2 - \sqrt{3}(63.57) = -67.906$$

$$\hat{b} = \bar{X} + \sqrt{3} S = 42.2 + \sqrt{3} (63.57) = 152.3064$$

Hence we can conclude that value of $\hat{b}$ is approximately very close to the sample value but $\hat{a}$ is very far from the true sample value, so it is the potential weakness of methods of moments.

(iv) $X \sim U(0, b)$

$$L = \prod_{i=1}^n f(x_i)$$

$$L = \left(\frac{1}{b}\right)^n$$

$$\log L = -n \log b$$

$$\frac{dL}{db} = -\frac{n}{b}$$

$$\frac{dL}{db} = 0, \text{ gives } b = \infty \text{ which is not possible}$$

Also,

$\hookrightarrow$

$$\frac{d^2L}{db^2} = \frac{n}{b^2}$$

$\frac{n}{b^2} > 0$, which we have a minimum likelihood estimate

Thus, we say, $\hat{b} = \max(X_i)$