

Assignment 1 [PSS]

$$Q1) \therefore \hat{\lambda} = \frac{\sum x_i}{n}$$

$$\hat{\lambda} = 200$$

$\therefore$ 95% CI of $\hat{\lambda}$

$$= \hat{\lambda} \pm Z_{2.5\%} \sqrt{\frac{\hat{\lambda}}{n}}$$

$$\therefore = 200 \pm 1.96 \times \sqrt{200}$$

$$[172.2814, 227.718]$$

$$Q2) i) E[\bar{x}] = E\left[\frac{\sum x}{n}\right]$$

$$V[\bar{x}] = V\left[\frac{\sum x}{n}\right]$$

$$= \frac{1}{n} \sum E[x]$$

$$= \frac{1}{n^2} \sum V[x]$$

$$= \frac{1 \cdot n \mu}{n} = \mu$$

$$= \frac{1 \cdot n \cdot \sigma^2}{n^2}$$

$$= \sigma^2 / n$$

$$ii) E(S^2) = E\left(\frac{1}{n-1} \sum (x - \bar{x})^2\right) \quad E[x^2] = V[x] + E[x]^2$$

$$= \frac{1}{n-1} E(\sum x^2 - n \bar{x}^2)$$

$$= \frac{1}{n-1} E(\sum (\mu^2 + \sigma^2) - n(\mu^2 + \frac{\sigma^2}{n}))$$

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

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

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

$V[S^2] = 2(n-1)$

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

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

$V[S^2] = \frac{2\sigma^4}{n-1}$

ii) $P[0.5\sigma^2 < S^2 < 0.5\sigma^2] \quad E[S^2] = \sigma^2$

$\therefore P[0.5 \times 100 < S^2 < 1.5 \times 100]$

$P[50 < S^2 < 150]$

$P\left[\frac{50 \times 9}{100} < \chi^2_9 < \frac{150 \times 9}{100}\right]$

$P[\chi^2_9 < 4.5] - P[\chi^2_9 < 13.5]$

$0.8587 - 0.1245$

$= 0.7342$

$$Q3) i) h = \frac{[P(X=0)]^{86} \cdot [P(X=1)]^{75} \cdot [P(X=2)]^{16}}{[P(X=3)]^2 \cdot [P(X=4)]^7}$$

$$\therefore = \left[\text{constant} \cdot p^0 \cdot (1-p)^4 \right]^{86} \times \left[4C_1 \times p \times (1-p)^3 \right]^{75} \\ \times \left[4C_2 \cdot p^2 \cdot (1-p)^2 \right]^{16} \cdot \left[4C_3 \cdot p^3 \cdot (1-p) \right]^2 \cdot \left[4C_4 \cdot p^4 \cdot (1-p)^0 \right]$$

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

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

$$\frac{d \log h}{dp} = \frac{117}{p} + \frac{603}{(1-p)} = 0$$

$$\hat{p} = \frac{117}{720} = 0.1625$$

iii) H_0 : ~~model~~ ~~all~~ claims are from binomial
 H_1 : are not from binomial

	0	1	2	3	4
O_i	86	75	16	2	1
E_i	88.542	68.724	19.998	2.574	0.108

$$T.S = \frac{\sum (O_i - E_i)^2}{E_i} \sim \chi^2_{5-1-1=3}$$

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

$\therefore$ At 5% LOS we have insufficient evidence to reject H_0 .

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

i] Comparing to PDF to Pareto Dist.

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

∴ β corresponds to λ

$$c = 3$$

$$E[x] = \frac{\lambda}{\alpha-1} = \frac{\beta}{2}$$

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

ii] ∴ $E[x] = \bar{x}$

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

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

$$\text{Bias} = E[\hat{\beta}] - \beta$$

$$= 2E[\bar{x}] - \beta$$

$$= \frac{2 \sum E[x_i]}{n} - \beta$$

$$= \frac{2 \cdot n \cdot \frac{\beta}{2}}{n} - \beta = 0$$

unbiased estimator

$$MSE = \text{Var}(\hat{\beta}) + \text{Bias}^2$$

$$= V[2\bar{x}] = \frac{4 \sum V[x]}{n^2}$$

∴ as $n \rightarrow \infty$

$$MSE \rightarrow 0$$

$$= \frac{4 \cdot n \cdot \frac{3\beta^2}{4}}{n^2}$$

It is consistent.

$$= \frac{3\beta^2}{n}$$

$$iii) \hat{\beta} = b\bar{x}$$

$$E[\hat{\beta}] = b E[\bar{x}]$$

$$= \frac{b}{n} \cdot n \cdot \frac{\beta}{2}$$

$$= \frac{b\beta}{2}$$

$$V[\hat{\beta}] = b^2 V[\bar{x}]$$

$$= \frac{b^2}{n^2} \cdot n \cdot \frac{3\beta^2}{4}$$

$$= \frac{3}{4} \frac{b^2 \beta^2}{n}$$

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

$$= \frac{3}{4} \frac{b^2 \beta^2}{n} + \left(\frac{b^2 \beta^2}{4} + 4\beta^2 - 4\beta^2 b \right)$$

$$M = \frac{\beta^2}{4n} (3b^2 + n b^2 + 4n - 4nb)$$

$$\frac{dM}{db} = \frac{\beta^2}{4n} (b + 2nb - 4n) = 0$$

$$b + 2nb - 4n = 0$$

$$\therefore b = \frac{2n}{3+n} = 2 / (3 + n/n)$$

$$iv) E(\hat{\beta}) = \frac{b\beta}{2}$$

$$= \frac{2}{(1+3/n)} \times \frac{\beta}{2}$$

$$= \frac{n\beta}{n+3}$$

$$\text{Bias} = E(\hat{\beta}) - \beta$$

$$= \frac{n\beta}{n+3} - \beta$$

$$= \frac{-3\beta}{n+3}$$

The estimator underestimates the parameter.

$$MSE = \frac{\beta^2}{4n} (3b^2 + nb^2 + 4n - 4nb)$$

$$= \beta^2 / 4n [b^2(n+3) + 4n(1-b)]$$

$$= \frac{\beta^2}{4n} \left[\frac{4n^2}{n+3} + 4n \frac{(3-n)}{n+3} \right]$$

$$MSE = \frac{12n\beta^2}{n+3}$$

$n \rightarrow \infty$. MSE doesn't tend to 0. It's inconsistent.

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

$$E[X] = \frac{a+b}{2}$$

$$V[X] = \frac{1}{12} (b-a)^2$$

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

$$= S^2 = \frac{1}{12} (b-a)^2$$

$$2\bar{X} - b = a \rightarrow \textcircled{1}$$

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

$$\sqrt{12S^2} = 2\bar{X} - 2a$$

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

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

iii) Sample observation

1, 3, 5, 7, 9

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

$$\bar{X} = 5$$

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

$$S = 3.16227$$

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

$$-0.4772$$

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

$$\therefore \hat{a} = 10.477$$

$$\hat{b} = 10.477$$

$\therefore \hat{b}$ is better estimator than $\hat{a}$.

$$\text{Ex) } h = \prod_{i=1}^n f(x_i) = \left(\frac{1}{b}\right)^n$$

$$\ln h = n \ln(1/b)$$

$$\frac{d \ln h}{d b} = \frac{n}{1/b} \cdot \frac{-1}{b^2}$$

$$= -\frac{n}{b} = 0$$

$$\text{Hence } \frac{d^2 \ln h}{d b^2} = \frac{n}{b^2} > 0$$

$$\therefore \hat{b} = \max(x_i)$$

$$16] H_0: p = 0.1$$

$$H_1: p > 0.1$$

$$a) P(\text{Type I error}) = P(\text{rejecting } H_0 \mid H_0 \text{ is true})$$

$$= P(X \geq 3 \mid p = 0.1) + P(X \geq 5 \mid 0.1) \dots$$

Public	24	45	29	33	20	40	26.5
Put	30	34.5	30	40	28.5	36	30.5
	25	27.5					
	30.5	35.5					

i) $n = 9$

$$\bar{X}_1 = 30$$

$$S_1^2 = 64.3125$$

$$\bar{X}_2 = 35.055$$

$$S_2^2 = 67.403$$

iii) $H_0: \mu_1 = \mu_2$
 $H_1: \mu_1 \neq \mu_2$

$$S_p^2 = \frac{1}{16} [514.5 + 606.827]$$

$$= 70.07$$

$$t_{\text{cal}} = \frac{(\bar{X}_1 - \bar{X}_2) - (\mu_1 - \mu_2)}{\sqrt{(70.0704)(0.2222)}} \sim t_{n_1+n_2-2}$$

$$= 0.00022246 \quad 1.3216$$

CI at 95% LOS = 2.12 ± 0.3246
 $(1.7954, 2.4446)$

$$t_{\text{cal}} < t_{\text{tab}}$$

We reject H_0 .

ii) $H_0: \mu_1 = \mu_2$

g) i) Unbiased estimator

This is when bias is 0. i.e. the estimator's expected value is equal to population parameter

ii) Biased

is the measure of difference between the expected value of the estimator and the parameter being estimated

iii) $MSE(\hat{\theta}) = \text{var}(\hat{\theta}) + \text{Bias}^2$

iv) The estimator having less MSE is more consistent

v) They both will be considered to be consistent when MSE minimizes such that $n \rightarrow \infty$ and $MSE \rightarrow 0$

vi) We can estimate θ by method of moments. We can also use a more preferred method of Maximum Likelihood Estimate.

97

i) The variable $\frac{W(0,1)}{\sqrt{t_k/K}} \sim t_k$ is defined as

$$\frac{W(0,1)}{\sqrt{t_k/K}} \sim t_k$$

where,

t_k is the t distribution with ' K ' degrees of freedom
 K is the number of samples or simply ' $n-1$ '.

ii) mean = 0

$$\text{variance} = \frac{K}{K-2}$$

iii) $n = 10$

$$\bar{x} = 50$$

$$s^2 = 48.667$$

$$a) \therefore \frac{\bar{x} - \mu}{s/\sqrt{n}} \sim t_{n-1}$$

$$P(t_{9,0.5\%} < \frac{\bar{x} - \mu}{s/\sqrt{n}} < t_{9,99.5\%}) = 0.99$$

$$TS = \bar{x} \pm 3.25 \times \sqrt{\frac{48.667}{10}} = 50 \pm 3.25(2.206)$$

$$\therefore 99\% CI = (42.83, 57.17)$$

iv) From (iii)

$$\frac{\bar{x} - \mu}{s/\sqrt{n}} \sim N(0, \sqrt{\frac{n-1}{n-3}})$$

$$= 50 - 2.58 \sqrt{\frac{48.667 \times 9}{17}}$$

$$\frac{\bar{x} - \mu}{s/\sqrt{n}} \sim N(0, \sqrt{9/17})$$

$$99\% CI = (43.54, 56.45)$$

$$T.S = \bar{x} - 2.05\% \sqrt{s^2/n}$$

10] $n = 1000$ $X \sim \text{Poi}(n\lambda)$
 $\lambda = 3$ $X \sim \text{Poi}(3000)$

$H_0 : n\lambda \geq 3000$

$H_1 : n\lambda \leq 3100$

i) $P(\text{Type 1 error}) = P(\text{Reject } H_0 \mid H_0 \text{ is True})$
 $= P(X > 3100 \mid X \sim \text{Poi}(3000))$

$P(X > 3100) = P(Z > \frac{3100 - 3000}{\sqrt{3000}})$

$= P(Z > 1.8257)$

$= \underline{\underline{0.0343}}$

ii) Type 2 error = event of accepting H_0 when false

iii) Power of test = $P(\text{reject } H_0 \mid H_0 \text{ is false})$

$P[X > 3100 \mid X \sim \text{Poi}(n\lambda) \approx N(n\lambda, n\lambda)]$

The value of power of test depends on the value of the parameter under hypothesis

iv) $\hat{u}$ is the estimated value

$\hat{u} \sim N(u, u/n)$ 99% CI

$\frac{\hat{u} - u}{\sqrt{\hat{u}/n}} \sim N(0,1) = (2.7613, 2.0387)$

$P(-2.578 < \frac{2900 - u}{\sqrt{2900/100}} < 2.578) = 0.99$

$$n = 12 \quad \sum x = 213200 \quad \sum x^2 = 4919860000$$

$$\bar{x} = 17767 \quad s = 10144$$

$$\text{new } \sum x_1 = 213200 - 11000 - 48000$$

$$+ 27500 + 31500$$

$$= 213200$$

$$\bar{x}_1 = \frac{213200}{12} = 17767$$

$$\sum x_2 = 424336000$$

$$s_1^2 = \frac{1}{n-1} [\sum x_1^2 - n(\bar{x}_1)^2]$$

$$= \frac{1}{11} (424336000 - 3787995468)$$

$$s^2 = 41396775.64$$

There is no change in $\bar{x}$. But s^2 decreases, showing that new data is better sample as the var has reduced.

$$112] \text{ i) } X \sim U(0, \theta)$$

The CRLB result holds under very general conditions except where the range of the dist. involves the parameter.

iii) Bias of estimator $\hat{\theta}(c)$ is given

$$\text{Bias}(\hat{\theta}(c)) = E[\hat{\theta}(c) - \theta]$$

$$= E[cY - \theta]$$

$$= c \cdot E[Y] - \theta$$

$$= c \cdot \frac{n\theta}{n+1} - \theta =$$

$$\left(\frac{cn}{n+1} - 1\right)\theta$$

$$MSE [\hat{\theta}(c)] = E[(\hat{\theta}(c) - \theta)^2]$$

$$= E[(cY - \theta)^2]$$

$$= c^2 E[Y^2] - 2c\theta E(Y) + \theta^2$$

$$= c^2 \cdot \frac{n\theta^2}{n+2} - 2c\theta \cdot \frac{n\theta}{n+1} + \theta^2$$

$$= \frac{c^2 \cdot n\theta^2}{n+2} - \frac{c \cdot 2n\theta^2}{n+1} + \theta^2$$

V) in order to minimize the MSE

$$0 = \frac{d}{dc} MSE[\hat{\theta}(c)]$$

$$c_m = \frac{n+2}{n+1} > 0 \therefore \text{It is minimum.}$$