

Probability Assignment

1. $X \sim \text{Pois}(\theta)$ $E(X) = \theta$ $V(\theta) = \theta$

$P[X \leq 200] = P[X \leq 200.5]$

$\rightarrow P\left[\frac{X - \theta}{\sqrt{\theta}} \leq \frac{200.5 - \theta}{\sqrt{\theta}}\right] \sim N(0,1)$

$\Rightarrow P[X = 200] \rightarrow P[199.5 < X < 200.5] \approx 0.$

$\rightarrow P[-1.96 < Z < 1.96]$

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

C.I: $\hat{\theta} = \theta \pm 1.96 \times \sqrt{\theta} = 200 \pm 1.96 \sqrt{200}$
 $= 200 \pm 1.96 \times 10\sqrt{2} = 200 \pm 27.7185$

$\rightarrow X_1 = \theta \pm 1.96\sqrt{\theta}$ C.I: $[172.28, 227.7185]$

2. $X_i \rightarrow i = 1, 2, 3, 4, \dots, n$ $E(X_i) = \mu$ $V(X_i) = \sigma^2$

i) $\bar{X} = \frac{\sum X_i}{n}$ $E(\bar{X}) = \frac{E(\sum X_i)}{n} = \frac{\sum E(X_i)}{n}$

$= \frac{\sum \mu}{n} = \frac{n \times \mu}{n} = \mu = E(\bar{X})$

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

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

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

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

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

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

$= \left(\frac{1}{n-1}\right) \times \sum E(X_i^2)$

$E(X_i^2) = V(X_i) + [E(X_i)]^2$ $E(\bar{X}^2) = V(\bar{X}) + [E(\bar{X})]^2$
 $= \sigma^2 + (\mu)^2$ $= \frac{\sigma^2}{n} + \mu^2$

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

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

$$E(S^2) = \sigma^2$$

$$\rightarrow X_i \sim N(\mu, \sigma^2)$$

$$\rightarrow V(S^2) = \frac{(n-1)\sigma^4}{\sigma^2} \sim \chi^2_{n-1}$$

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

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

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

$$\text{iv.) } P[E(S^2) = \frac{(n-1)\sigma^2}{(n-1)}] = \frac{(n-1)\sigma^2}{(n-1)}$$

$$\therefore E(S^2) = \sigma^2$$

$$\rightarrow P[\chi_1 < \sigma^2 < \chi_2] = 0.50$$

$$\therefore P[\chi_1 < \sigma^2 < \chi_2] = (1 - 50\%) = 0.50$$

$$\rightarrow Z_{25\%} \rightarrow 0.6745$$

$$\star \rightarrow P\left[\frac{1}{\chi_1} < \frac{1}{\sigma^2} < \frac{1}{\chi_2}\right] = 0.50$$

$$\rightarrow P\left[\frac{(n-1)S^2}{\chi_1} < \frac{(n-1)S^2}{\sigma^2} < \frac{(n-1)S^2}{\chi_2}\right] = 0.50$$

$$\rightarrow \frac{(n-1)S^2}{\chi_1} = \chi^2_{n-1, 25\%} \quad \left| \quad \frac{(n-1)S^2}{\chi_2} = \chi^2_{n-1, 75\%}\right.$$

$$\rightarrow \therefore \chi_1 = \frac{9 \times}{\chi^2_{9, 25\%}} \quad \chi_2 = \frac{9 \times}{(8.343)}$$

$$\rightarrow \frac{(n-1)S^2}{\sigma^2} \sim \chi^2_{n-1} \quad S^2 = \frac{\chi^2_{9, 30\%} \times (100)}{9} = 92.7$$

$$P\left[\frac{(n-1)S^2}{\sigma^2} \sim \chi^2_{n-1}\right]$$

$$\text{iv.) } \sigma^2 = 100$$

$$E(S^2) = \sigma^2 = 100$$

$$\rightarrow P[50 < E(S^2) < 150] = P[E(S^2) < 150] - P[E(S^2) < 50]$$

$$\rightarrow P[\sigma^2 < 150] - P[\sigma^2 < 50]$$

$$= P\left[\frac{9 \times S^2}{100} < \frac{9 \times 150}{100}\right] - P\left[\frac{9 \times S^2}{100} < \frac{50 \times 9}{100}\right]$$

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

$$= 0.8587 - 0.1245$$

$$= 0.7342$$

$$N = 4$$

Q.3	$X:$	0	1	2	3	4
	$P(X=x)$	q^4	$p^3 q^3 \cdot 4$	$6 p^2 q^2$	$4 p^3 q^3$	p^4
	Freq:	86	75	16	2	1

$$L = \prod_{i=1}^n f(x_i) = (q^4)^{86} \cdot (p^3 q^3 \cdot 4)^{75} \cdot (6 p^2 q^2)^{16} \cdot (4 p^3 q^3)^2 \cdot (p^4)^1$$

$$L = (1-p)^{344} \cdot (1-p)^{225} \cdot p^{75} \cdot p^{32} \cdot (1-p)^{32} \cdot p^6 \cdot (1-p)^2 \cdot p^4 \cdot \text{const}$$

$$\ln L = \ln [(1-p)^{603} \cdot p^{117} \cdot \text{const}]$$

$$\ln L = 603 \ln(1-p) + 117 \ln(p) \quad \text{Divide by } 1-p$$

$$\therefore \frac{117}{p} = \frac{603}{1-p} \quad \therefore \frac{1}{p} - 1 = \frac{603}{117} \quad \therefore \frac{1}{p} = 6.15$$

$$\hat{p} = 0.1625$$

ii.) Obs: 0 1 2 3 4

Exp freq: $P(X=0) \cdot \text{Total} = (1-0.1625)^4 \times 180 = 88.55$

$P(X=1) \cdot T \rightarrow 68.729$

$P(X=3) \rightarrow 2.587$

$P(X=2) \rightarrow 20$

$P(X=4) \rightarrow 0.1255$

O	86	75	16	2	1
E	88.55	68.72	25.87	2.587	0.1255
			20.20		

$$\frac{(O-E)^2}{E} : 0.0734 \quad 0.573$$

As Exp freq of $X=3$ & 4 less than 5, we add it to 20.

$\therefore O \quad 86 \quad 75 \quad 19$

$E \quad 88.55 \quad 68.72 \quad 22.7125$

$$\frac{(O-E)^2}{E} : 0.0734 \quad 0.573 \quad 0.606 = 1.2532$$

$$\sum \frac{(O-E)^2}{E} \sim \chi^2_{3-1-1} = \chi^2_1 = 1.2532$$

$\rightarrow \chi^2_1 = 1.2532, \chi^2_1 \rightarrow \text{at } 5\% : 3.841$

$\text{at } 10\% : 2.706$

As this is a fit test.

As $\chi^2_{\text{cal}} < \chi^2_{\text{tab}}$

Do not Reject H_0

Q.4 $f(x) = \frac{c\beta^3}{(x+\beta)^4} \quad x > 0, \beta > 0$

i. $E(x) = \int_0^{\infty} x \cdot f(x) dx = \int_0^{\infty} x \cdot \frac{c\beta^3}{(x+\beta)^4} dx$

By looking at Pareto Distribution,

$\rightarrow f(x) = \frac{c\beta^3}{(x+\beta)^4} = \frac{\alpha \lambda^\alpha}{(\lambda+x)^{\alpha+1}} \therefore \lambda = \beta, \alpha = 3, c = 3$

$\rightarrow \alpha = 3, \beta = \lambda, E(x) = \frac{\lambda}{\alpha-1} = \frac{\beta}{2} = E(x)$

$V(x) = \frac{\alpha \lambda^2}{(\alpha-1)^2(\alpha-2)} \rightarrow \frac{3 \times \beta^2}{(2)^2 \cdot (1)} = \frac{3 \times \beta^2}{4} = \text{Var}(x)$

ii) By MOM: $E(x) = \bar{x}$

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Random Sample $\rightarrow X_1, X_2, X_3, \dots, X_n$

$\bar{X} = \frac{\sum_{i=1}^n X_i}{n} = \bar{X}_n \quad 1, 2, 3, 4, 5 \rightarrow U(1, 5)$

$E(x) = \frac{\beta}{2} \therefore \hat{\beta} = 2\bar{X}_n = 3$

$a = 3 - 2/\sqrt{3} \cdot \sqrt{3} = 1$

iii) $E(\hat{\beta}) = E(2\bar{X}_n) = 2 \cdot \frac{3}{2} = 3$

$b = 3 + 2/\sqrt{3} \cdot \sqrt{3} = 5$

$= 2 \sum_{i=1}^n E(X_i) = 2 \times \frac{\beta}{2} = \beta = \text{Estimator}$

As $E(g(\theta)) = \theta$, it is unbiased.

$\rightarrow \text{Consistency} = \text{MSE} = \text{Var} + \text{Bias}^2 = \frac{3 \times \beta^2}{4} = 0.75 \beta^2$
for More Consistent

iii) MSE of $(b\bar{X}_n) \rightarrow \text{Estimator}$

Q5. $E(X) \sim \text{Unif}(a, b) \rightarrow E(X) = \bar{X}$

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

$$\frac{(b-a)^2}{12} = s^2 \rightarrow (2\bar{X} - 2a)^2 = 12s^2$$

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

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

$$\text{ii)} \frac{(b - 2\bar{X} + b)^2}{12} = s^2 \quad \therefore [2(b - \bar{X})]^2 = 12s^2$$

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

$$\therefore b = \bar{X} + \sqrt{3}s$$

(a)

iii) Sample: 1, 2, 3, 4, 5 $\rightarrow U(1, 5)$

$$\bar{X} = \frac{a+b}{2} = \frac{5+1}{2} = 3 \quad s^2 = \frac{(b-a)^2}{12} = \frac{16}{12} = \frac{4}{3} \quad s = 2/\sqrt{3}$$

$$\rightarrow a = \bar{X} - \sqrt{3}s = 3 - \sqrt{3} \times 2/\sqrt{3} = 1 \quad s = 1.581$$

$$b = \bar{X} + \sqrt{3}s = 3 + \sqrt{3} \times 2/\sqrt{3} = 5$$

$$\rightarrow a = 3 - \sqrt{3} \times (1.581) = 0.2616 \quad b = 3 + \sqrt{3} \times (1.581) = 5.7383$$

$\rightarrow$ As a and B values are not same, H0H is weak.

iv. PDF of Cont: $\frac{1}{b-a}$

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

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$$L = \left(\frac{1}{b-a}\right)^N = \left(\frac{1}{b}\right)^N \quad \ln L = N \cdot \ln\left(\frac{1}{b}\right)$$

$$\frac{d \ln L}{db} = \frac{N \times -1}{(1/b) \cdot b^2} = \frac{b \times N (-1)}{b^2} = -\frac{N}{b} = 0$$

$$\rightarrow \ln L = N \cdot \ln\left[\frac{1}{b-a}\right] = \frac{d \ln L}{d(b)} \cdot \frac{N}{(1/(b-a))} \times \frac{-1}{(b-a)^2}$$

$$\rightarrow \text{when } U(0, b), E(x) = \frac{1}{b-a} = \frac{1}{b}$$

26 $H_0: p = 0.1$ $H_1: p > 0.1$

i) $n = 12 \rightarrow$ first step: If 3 or more hits i.e.

- Hits: If $p \geq 0.25$, reject H_0

$\rightarrow P[N \geq 3 / p = 0.1] = 1 - P[N = 0, N = 1, N = 2 / p = 0.1]$

$= 1 - [{}^{12}C_0 \cdot (0.1)^0 \cdot (0.9)^{12} + {}^{12}C_1 \cdot (0.1)^1 \cdot (0.9)^{11} + {}^{12}C_2 \cdot (0.1)^2 \cdot (0.9)^{10}]$
 $= 1 - (0.282429 + 0.3765 + 0.2301) = 0.1109$

$\rightarrow n = 24, 1 - P[N = 0, N = 1, N = 2, N = 3, N = 4 / p = 0.1]$

iii) $P[N > P[N < 5 / p = 0.3]]$

$= P[N = 0, N = 1, N = 2, N = 3, N = 4 / p = 0.3]$

$\rightarrow {}^{24}C_0 \cdot (0.3)^0 \cdot (0.7)^{24} + {}^{24}C_1 \cdot 0.3 \cdot (0.7)^{23} + {}^{24}C_2 \cdot (0.3)^2 \cdot (0.7)^{22} +$
 ${}^{24}C_3 \cdot (0.3)^3 \cdot (0.7)^{21} + {}^{24}C_4 \cdot (0.3)^4 \cdot (0.7)^{20}]$

$\rightarrow 0.0001915 + 0.00197 + 0.009711 + 0.030 + 0.0686$
 $= 0.11055$

iv) $p = ?$ $\text{Bin}(12, 1/3) \sim N(p, \frac{(1-p)(p)}{n})$

$\rightarrow \chi_1 \rightarrow$

Lower Bound -? $Z_{5\%} \rightarrow 1/3 \pm 1.6449 \times (0.14907)$

$\therefore 0.333 + 0.245, 0.333 - 0.245$

$\rightarrow 0.578, 0.0877$

1.2816

v) $TS: 0.33 - 0.278 = 0.052$ $\rightarrow Z_{10\%} = 1.6449$

$\sqrt{\frac{0.33(0.667)}{10}}$

$\therefore Z_{cal} < Z_{tab},$

we DN Reject H_0

As TS lied in the CI, we accepted H_0

vi) $\mu_A = \mu_B = H_0$ $\mu_A \neq \mu_B = H_1$ $\sigma = \mu_A - \mu_B = 0$

TS: $\frac{(\mu_A - \mu_B) - \sigma}{\sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}} \sim N(0, 1) \rightarrow \frac{(70 - 65) - 0}{\sqrt{\frac{54^2}{12} + \frac{70^2}{15}}}$

$\rightarrow 5 = 0.2094 \rightarrow \text{At } 2.5\% \rightarrow 1.96$

$\text{At } 5\% \rightarrow 1.6449$

$\sqrt{243 + 326.6}$

As $Z_{cal} < Z_{tab}$, we DN H_0 for 5% LOS.

Assum $\rightarrow$ This SD is popⁿ SD.

Q7i) Public $\rightarrow \bar{x} = 30$ $\sigma^2 = 57.16667$ Private: $\bar{x} = 35.05$ $\sigma^2 = 67.4027$

ii.)

Q8: i.) Unbias Estimator: Difference between expected value of estimator and estimator is 0. $E(\hat{\theta}) = \theta$ If $\hat{\theta}$ is used to estimate θ , then if both are same, then $\hat{\theta}$ is an unbiased estimator.ii) Bias: when $\hat{\theta}$ used to estimate θ , does not give value same as θ , then is called bias.iii) MSE: Variance of Estimator w.r.t. true parameter value, $MSE = \text{Variance} + \text{Bias}^2$

iv) As MSE of $\tilde{\theta}$ is less, it is more consistent.
MSE is variance of estimator. if estimator is less, then values are more consistent.

v) They would be termed consistent, when MSE $\downarrow$ or gradually becomes zero.

vi) Method of Moments of Method of Likelihood Estimator.

8.9 i) In real life, population variance is never known, so we use 'S' as sample variance.

If $Z \sim N(0,1)$ and $C \sim \chi^2_k$, then $\frac{Z}{\sqrt{C/k}} \sim t_k$

i.e. the ratio of Standard Normal Distribution (Z) and χ^2 distribution with 'k' degree of freedom (C)'s root follows 't' distribution with 'k' degree of freedom.

ii) when $k > 2$, for 't': Mean = 0
Variance = $\frac{k}{k-2}$

iii) $n=10$ $\mu=50$ $S=48.667$ $\alpha=1\% \therefore \alpha/2=0.5\%$

$$a) CI = \bar{x} \pm t_{\alpha/2, n-1} \times \left(\frac{S}{\sqrt{n}} \right)$$

$$= 50 \pm t_{0.5\%, 9} \times \frac{\sqrt{48.667}}{\sqrt{10}}$$

$$x_1 = 50 + 3.250 \times \frac{2.206}{\sqrt{10}} = 57.1696$$

$$x_2 = 50 - 3.250 \times \frac{2.206}{\sqrt{10}} = 42.8305$$

$\rightarrow$ CI for 99% is 42.8305 & 57.1696

b) Mean = 0 Var = $k/k-2$ $k=9$ $t_k = t_{n-1}$

$$Var = 9/9-2 = 1.285714$$

$$\rightarrow P[x_1 < \bar{x} < x_2] = 0.99$$

$$P\left[\frac{x_1 - \mu}{\sigma/\sqrt{n}} < z = \frac{\bar{x} - \mu}{\sigma/\sqrt{n}} < \frac{x_2 - \mu}{\sigma/\sqrt{n}} \right] = 0.99$$

$$z_{\alpha/2} = z_{0.5\%}$$

$$\therefore x_1 = \mu + 2.5758 \times \frac{\sigma^2}{\sqrt{n}}$$

$$x_2 = 0.38117$$

$$x_2 = \mu - 2.5758 \times \frac{\sigma^2}{\sqrt{n}}$$

$$x_1 = -0.38117$$

Q.11 $n=12$ $\sum x = 218200$ $\sum x^2 = 4919860000$

Sample Mean = 17,767 Sample Var SD = 10,144

i) Sample Mean = $\frac{\sum x}{N}$

ii) Sample SD = $\sqrt{\frac{\sum x^2 - n(\bar{x})^2}{n-1}}$

Sample Mean = 17767 = $\frac{\sum x_s}{12}$ $\therefore \sum x_s = 213204$

New $\sum x_s = 213204 - 11000 - 48000 + 27500 + 31500$
 $= 213204$

New Sample Mean = $\frac{\text{New } \sum x_s}{N} = \frac{213204}{12} = 17,767$

ii) Sample SD = 10,144 = $\sqrt{\frac{\sum x_s^2 - n(\bar{x}_s)^2}{n-1}}$
 $= \frac{(102906736) - (4919860000) - 12 [17767]^2}{11}$

$\rightarrow \therefore \sum x_s^2 = 4919903564$

New $\sum x_s^2 = 4919903564 - (11000)^2 - (48000)^2$
 $+ (27500)^2 + (31500)^2$

New = 4243403564

$\therefore$ New SD = $\sqrt{\frac{\text{New } \sum x_s^2 - (n) \cdot (\text{New Mean})^2}{(n-1)}}$

Var_s = $\sqrt{41400736}$

$\rightarrow$ New Sample SD = 6434.34034

* With change inputs the new sample mean remained constant, while new sample deviation reduced.

Q12 i) CRLB is used when $n \rightarrow \infty$. Here in uniform distribution it ranges from 0 to θ , where θ is a parameter, doesn't tend to infinity

ii) $y = \max X_i$ $X_i \sim \text{Unif}(0, \theta)$ $\therefore Y$ also follows $(0, \theta)$

$$\rightarrow \text{Bias: } E(b\bar{X}_n) = b \times \sum E(X_i) = b \times \sum \frac{\beta \times 1}{2 \times n}$$

$$= b \times \cancel{n} \times \beta \times \frac{1}{\cancel{n}} = \frac{b \times \beta}{2} \quad n$$

$$\rightarrow \text{Bias: } \frac{b \times \beta}{2} - \beta = \frac{(b-2)\beta}{2} = \underline{\text{Bias}}$$

$$\text{Var}(\beta) = V(b\bar{X}_n) = b^2 \times \sum V(X_i) = \frac{b^2 \times n \times \frac{3}{4} \times \beta^2}{n^2} = \frac{3b^2\beta^2}{4n}$$

$$\rightarrow V(\beta) = \frac{3b^2\beta^2}{4n} \quad \text{Bias} = \frac{(b-2)\beta}{2}$$

$$\text{MSE} = \text{Var} - \text{Bias}^2 = \frac{3}{4n} b^2 \beta^2 - \frac{\beta^2}{4} [b^2 + 4 - 2b]$$

$$= \frac{3}{4n} b^2 \beta^2 - \frac{\beta^2 b^2}{4} + 4\beta^2 + 8b\beta^2$$