

True standard deviation ( $\sigma$ ) = 0.75

Samples taken follow a normal distribution.

(a) As we need to compute a 95% - CI,

$$\therefore \gamma = 0.95$$

$$n = 20$$

$$\text{Sample mean } (\bar{x}_n) = \underline{\underline{4.85}}$$

To compute 95% CI for the true average porosity of a certain seam, we need the following formulations,

$$\bar{x}_n - \frac{\sigma}{\sqrt{n}} q(1-\alpha_2) \leq \mu \leq \bar{x}_n - \frac{\sigma}{\sqrt{n}} q(\alpha_1) \quad (1)$$

where,

$$\alpha_1 + \alpha_2 = 1 - \gamma \quad (2)$$

As we will consider this to be a balanced CI,

$$\therefore \alpha_1 = \alpha_2$$

$$\therefore \alpha_1 + \alpha_2 = 1 - 0.95 = 0.05$$

$$2\alpha_1 = 0.05 \Rightarrow \alpha_1 = 0.025$$

$$\therefore \alpha_1 = \alpha_2, \quad \therefore \alpha_1 = \alpha_2 = 0.025$$

Considering the (1)st equation above and putting in the values,

$$4.85 - \frac{0.75}{\sqrt{20}} q(1-0.025) \leq \mu \leq 4.85 - \frac{0.75}{\sqrt{20}} q(0.025)$$

Solving this,

and considering the normal distribution table,

$$q_{\frac{1}{2}-0.025} = q_{0.975} = \underline{\underline{1.96}}$$

$$\& q_{0.025} = \underline{\underline{-1.96}}$$

$\Rightarrow$

$$4.85 - \frac{0.75}{\sqrt{20}} \times (1.96) \leq \mu \leq 4.85 - \frac{0.75}{\sqrt{20}} \times (-1.96)$$

$$\boxed{4.52 \leq \mu \leq 5.18}$$

This is the 95% CI for  $\mu \Rightarrow \underline{\underline{[4.52, 5.18]}}$   
Anw.

(b)

Now, to compute the 98% CI for true average porosity of another seam,

$$\boxed{\gamma = 0.98}$$

$n=16$ , sample average porosity ( $\bar{x}_n$ ) = 4.56.

$$\underline{\underline{\sigma = 0.75}}$$

Considering the ①st and ②nd equation in part (a),  
when we are considering a balanced CI here,

$$\alpha_1 + \alpha_2 = 1 - 0.98 = 0.02$$

$$\left[ \because \alpha_1 = \alpha_2 \right]$$

$$\therefore 2\alpha_1 = 0.02 \Rightarrow \boxed{\alpha_1 = 0.01}$$

$$\left[ \therefore \alpha_1 = \alpha_2 = 0.01 \right]$$

$$4.56 - \frac{0.75}{\sqrt{16}} q(1-0.01) \leq \mu \leq 4.56 - \frac{0.75}{\sqrt{16}} q(0.01)$$

From the normal distribution table,

$$q(1-0.01) = q(0.99) = \underline{\underline{2.33}}$$

$$\therefore q(0.01) = -\underline{\underline{2.33}}$$

$\Rightarrow$

$$4.56 - \frac{0.75}{\sqrt{16}} \times (2.33) \leq \mu \leq 4.56 - \frac{0.75}{\sqrt{16}} \times (-2.33)$$

$$4.124 \leq \mu \leq 4.996$$

$$4.12 \leq \mu \leq 5.0$$

This is the 98% CI for  $\mu \Rightarrow \underline{\underline{[4.12, 5.0]}}$  Ans.

(c) for a 95% interval,  $\underline{\underline{\gamma = 0.95}}$

$$\underline{\underline{\sigma = 0.75}}.$$

$$\text{Width of the interval (L)} = \underline{\underline{0.40}}$$

To find how large the sample size ( $n$ ) is  $\Rightarrow$   
from the previous parts, we know that the confidence  
intervals are given by:-

$$\left( \bar{x}_n - \frac{\sigma}{\sqrt{n}} q(N(0,1), 1-\alpha_2), \bar{x}_n - \frac{\sigma}{\sqrt{n}} q(N(0,1), \alpha_1) \right)$$

$\therefore$  the width of the interval can be computed by  $\Rightarrow$

$$L = \left[ \bar{X}_n - \frac{\sigma}{\sqrt{n}} q(N(0,1), \alpha_1) \right] - \left[ \bar{X}_n - \frac{\sigma}{\sqrt{n}} q(N(0,1), 1-\alpha_2) \right]$$

$\therefore$  we will consider this as a balanced CI,  
 $(\alpha_1 = \alpha_2)$

$$\alpha_1 + \alpha_2 = 1 - 0.95 = 0.05$$

$$\Rightarrow 2\alpha_1 = 0.05 \Rightarrow \alpha_1 = 0.025$$

$$\Rightarrow \boxed{\alpha_1 = \alpha_2 = 0.025}$$

Putting in all the values in the ③rd equation above,  
& solve.

$$0.40 = \left[ \bar{X}_n - \bar{X}_n - \frac{\sigma}{\sqrt{n}} q(N(0,1), \alpha_1) + \frac{\sigma}{\sqrt{n}} q(N(0,1), 1-\alpha_2) \right]$$

$$0.40 = \frac{\sigma}{\sqrt{n}} \left[ q(N(0,1), 1-\alpha_2) - q(N(0,1), \alpha_1) \right]$$

$$\Rightarrow \sqrt{n} = \frac{0.75}{0.40} \left[ q(N(0,1), 1-\alpha_2) - q(N(0,1), \alpha_1) \right]$$

$$q(N(0,1), 1-\alpha_2) = q(N(0,1), 1-0.025)$$

$$\Rightarrow q(N(0,1), 0.975) = \underline{\underline{1.96}}$$

$$\therefore q(N(0,1), 0.975) = -q(N(0,1), 0.025)$$

$$\therefore q(N(0,1), 0.025) = -\underline{\underline{1.96}}$$

$$\Rightarrow \sqrt{n} = \frac{0.75}{0.40} \left[ 1.96 - (-1.96) \right] = \frac{0.75}{0.40} \times 3.92$$

$$n = \left( \frac{0.75}{0.40} \times 3.92 \right)^2 = 54.0225$$

Rounding up the value of  $n$

$$\boxed{n \approx 55}$$

$\therefore$  The sample size ( $n$ ) should be 55.

Ans.

(d)

for a 99% confidence interval,

$$\underline{\gamma = 0.99}, \quad \underline{\sigma = 0.75}$$

As we need to find the sample size necessary to estimate true average porosity ( $\mu$ ) to be within 0.2.

As this is a balanced CI for a normal distribution, it will be the same on both the sides from  $\mu$ ,

$$\therefore L = 2 \times 0.2 = 0.4$$

Again, this is a balanced CI ( $\alpha_1 = \alpha_2$ ),

$$\alpha_1 + \alpha_2 = 1 - \gamma = 1 - 0.99 = 0.01$$

$$\Rightarrow 2\alpha_1 = 0.01 \Rightarrow \boxed{\alpha_1 = 0.005}$$

$$\Rightarrow \boxed{\alpha_1 = \alpha_2 = 0.005}$$

Using the equation (3) in part (d),

$$0.4 = \cancel{0.75} \left[ \bar{X}_n - \bar{X}_n - \frac{\sigma}{\sqrt{n}} q(N(0,1), \alpha_1) + \frac{\sigma}{\sqrt{n}} q(N(0,1), 1 - \alpha_2) \right]$$

$$0.4 = \frac{0.75}{\sqrt{n}} \left[ q_{N(0,1), 1-0.005} - q_{N(0,1), 0.005} \right]$$

$$\Rightarrow 0.4 = \frac{0.75}{\sqrt{n}} \left[ q_{N(0,1), 0.995} - q_{N(0,1), 0.005} \right]$$

$$\Rightarrow 0.4 = \frac{0.75}{\sqrt{n}} [2.58 - (-2.58)]$$

$$\left[ \begin{array}{l} \therefore q_{N(0,1), 0.995} = 2.58 \\ q_{N(0,1), 0.995} = -q_{N(0,1), 0.005} \\ \therefore q_{N(0,1), 0.005} = -2.58 \end{array} \right]$$

$$\Rightarrow \sqrt{n} = \frac{0.75}{0.4} \times 2 \times 2.58$$

$$\Rightarrow n = \left( \frac{0.75}{0.4} \times 2 \times 2.58 \right)^2 = \underline{\underline{93.61}}$$

We will round up this value,

$$\boxed{n \approx 94}$$

Ans.

2. Sample mean compressive strength ( $\bar{x}_n$ ) = 64.41 MPa  
 Sample standard deviation ( $s_n$ ) = 10.32 MPa  
n = 18 tested specimen.

The distribution for compressive strength was found to be normal.

a) Confidence level ( $\gamma$ ) = 0.98 (98%)

To calculate the confidence interval for the true average compressive strength, we will go through the steps 1-4 as taught in the class.

Step 1:- find a pivot for the unknown parameters,  
 Here we have 2 unknown parameters, both,  $\mu$  and  $\sigma^2$  (mean and variance).

$$E(X_1, \dots, X_n; \mu, \sigma^2) = \frac{\bar{X}_n - \mu}{S_n/\sqrt{n}} \quad \text{where,}$$

$\bar{X}_n$  is sample mean,  $S_n$  is sample variance and  $\mu$  is the true mean/ average.

This is considered here ( $S_n$  instead of  $\sigma^2$ ) because here we have both  $\mu$  and  $\sigma$  unknown and  $\bar{X}_n$  and  $S_n$  have been provided.

Step(2):- To see the distribution of the pivot,  
 $\therefore$  we are concerned with true average,  
the distribution of the pivot ( $\bar{X}$ ) is  $t_{n-1}$  distribution.  
( $t$ -distribution with  $n-1$  degrees of freedom).  
(As shown in class),

$$\boxed{\bar{X} (x_1, \dots, x_n; \mu) = \frac{\bar{x}_n - \mu}{S_n / \sqrt{n}} \sim t_{n-1}}$$

As we can see above that the pivot is dependent on  $\mu$  but the distribution of the pivot is not!

Also,

Let  $q_\gamma$  be its quantile function,

$q_\gamma(x)$  = The value of  $u$  such that

$$\boxed{P(t_{n-1} \leq x) = \gamma}$$

Step(3):- Since we need to calculate the confidence interval with an upper and a lower bound, we will choose some plausible values for  $\alpha_1$  and  $\alpha_2$ .

We know:-

$$\boxed{\alpha_1 + \alpha_2 = 1 - \gamma}$$

$$\Rightarrow P(q(t_{n-1}, \alpha_1) \leq \bar{X}(x_1, \dots, x_n; \mu, \sigma^2) \leq q(t_{n-1}, 1 - \alpha_2)) = \gamma$$

Substituting the value of the pivot in the equation above:-

$$P\left(q_{\gamma}(t_{n-1}, \alpha_1) \leq \frac{\bar{x}_n - \mu}{S_n / \sqrt{n}} \leq q_{\gamma}(t_{n-1}, 1-\alpha_2)\right) = \gamma$$

Step(4):-

Solving the inequality for  $\mu$ , we get:-

Multiplying both sides by  $S_n / \sqrt{n}$  :-

$\Rightarrow$

$$P\left(\frac{S_n}{\sqrt{n}} q_{\gamma}(t_{n-1}, \alpha_1) \leq \bar{x}_n - \mu \leq \frac{S_n}{\sqrt{n}} q_{\gamma}(t_{n-1}, 1-\alpha_2)\right) = \gamma$$

Subtracting  $\bar{x}_n$  from ~~both~~ both the sides of the inequality,

$$P\left(\frac{S_n}{\sqrt{n}} q_{\gamma}(t_{n-1}, \alpha_1) - \bar{x}_n \leq -\mu \leq \frac{S_n}{\sqrt{n}} q_{\gamma}(t_{n-1}, 1-\alpha_2) - \bar{x}_n\right) = \gamma$$

Multiplying the inequality by  $-1$ , we get:-

$$P\left(\bar{x}_n - \frac{S_n}{\sqrt{n}} q_{\gamma}(t_{n-1}, 1-\alpha_2) \leq \mu \leq \bar{x}_n - \frac{S_n}{\sqrt{n}} q_{\gamma}(t_{n-1}, \alpha_1)\right) = \gamma$$

Thus, the confidence interval for  $\mu$  can be given as:-

$$\boxed{\bar{x}_n - \frac{S_n}{\sqrt{n}} q_{\gamma}(t_{n-1}, 1-\alpha_2) \leq \mu \leq \bar{x}_n - \frac{S_n}{\sqrt{n}} q_{\gamma}(t_{n-1}, \alpha_1)}$$

(1)

$\therefore$  confidence level is 98%,  $\boxed{\gamma = 0.98}$

Now, we will put in all the values in equation ① above,

Also,  $\because$  we know:-

$$\alpha_1 + \alpha_2 = 1 - \gamma = 1 - 0.98 = 0.02$$

$\therefore$  this is a balanced CI ( $\alpha_1 = \alpha_2$ ),

$$2\alpha_1 = 0.02 \Rightarrow \alpha_1 = 0.01$$

$$\Rightarrow \boxed{\alpha_1 = \alpha_2 = 0.01}$$

Now, from equation ①,

$$64.41 - \frac{10.32}{\sqrt{18}} q_{(t_{n-1}, 1-0.01)} \leq \mu \leq 64.41 - \frac{10.32}{\sqrt{18}}$$

$$q_{(t_{n-1}, 0.01)}$$

$\therefore$  we know,

$$q_{(t_{n-1}, 0.99)} = \underline{\underline{2.567}}$$

$$\& q_{(t_{n-1}, 0.99)} = - q_{(t_{n-1}, 0.01)}$$

$$\therefore q_{(t_{n-1}, 0.01)} = \underline{\underline{-2.567}}$$

$\Rightarrow$

$$64.41 - \frac{10.32}{\sqrt{18}} \times 2.567 \leq \mu \leq 64.41 - \frac{10.32}{\sqrt{18}} \times (-2.567)$$

$$58.16 \leq \mu \leq 70.654$$

$$\boxed{58.16 \leq \mu \leq 70.65} \quad \text{Ans}$$

$\therefore$  The confidence interval is  $[58.16, 70.65]$

Ans.

b)

Since, now we consider  $\gamma = 0.99$

as 99% CI of the form  $(-\infty, R_n)$ .

(Upper-bounded)

Also, we know that:-

$$\alpha_1 + \alpha_2 = 1 - \gamma$$

$$\Rightarrow \alpha_1 + \alpha_2 = 1 - 0.99 = 0.01 \Rightarrow \boxed{\alpha_1 + \alpha_2 = 0.01} \quad (2)$$

In order to get the interval as  $(-\infty, R_n)$ , the lower bound of the CI should be  $-\infty$ .

Considering the equation ① from the part a),

$$\bar{x}_n - \frac{s_n}{\sqrt{n}} q(t_{n-1}, 1 - \alpha_2) = -\infty \quad (\because \bar{x}_n \text{ is finite})$$

$$\Rightarrow \frac{s_n}{\sqrt{n}} q(t_{n-1}, 1 - \alpha_2) = \infty$$

$\therefore s_n$  and  $n$  is finite,  $\therefore q(t_{n-1}, 1 - \alpha_2) = \infty$

$\therefore$  we know that  $\boxed{q(1) = \infty}$ .

$$\therefore 1 - \alpha_2 = 1 \Rightarrow \boxed{\alpha_2 = 0}$$

Putting this in equation ② above,

$$\alpha_1 + 0 = 0.99 \Rightarrow \boxed{\alpha_1 = 0.99}$$

Putting back the lower bound of  $\infty$  and other values in RHS of the inequality ①,

$$-\infty \leq \mu \leq 64.41 - \frac{10.32}{\sqrt{18}} g_r(t_{n-1}, 0.01)$$

$$-\infty \leq \mu \leq 64.41 - \frac{10.32}{\sqrt{18}} \times (-2.567)$$

$$\left[ \because g_r(t_{n-1}, 0.01) = -2.567 \right]$$

$\Rightarrow$

$$-\infty \leq \mu \leq 70.654$$

$$\Rightarrow \boxed{-\infty < \mu \leq 70.65}$$

Ans.

$\therefore$  The required CI is  $\Rightarrow (-\infty, 70.65)$

where  $R_n$  comes out to be  $\frac{70.65}{18}$

Ans.

3.

Let  $X_1, X_2, \dots, X_n$  be a random sample from a uniform distribution on the interval  $[0, \Theta]$ , so, that:-

$$f(x) = \begin{cases} \frac{1}{\Theta}, & 0 \leq x \leq \Theta \\ 0, & \text{otherwise} \end{cases}$$

$\therefore Y = \min(X_i)$ ,

Let  $U$  be a random variable (rv) such that  $U = Y/\Theta$ ,

$$f_U(u) = \begin{cases} n u^{n-1}, & 0 \leq u \leq 1 \\ 0, & \text{otherwise} \end{cases}$$

(@) first, let's consider the LHS of the inequality:-

$$P\left(\left(\frac{\alpha}{2}\right)^{\frac{1}{n}} < U \leq \left(1 - \frac{\alpha}{2}\right)^{\frac{1}{n}}\right)$$

$$\left[\because U = Y/\Theta\right]$$

$$= P\left(\left(\frac{\alpha}{2}\right)^{\frac{1}{n}} < U \leq \left(1 - \frac{\alpha}{2}\right)^{\frac{1}{n}}\right)$$

$$= F\left(\left(1 - \frac{\alpha}{2}\right)^{\frac{1}{n}}\right) - F\left(\left(\frac{\alpha}{2}\right)^{\frac{1}{n}}\right) \quad \{ F \text{ is the cdf}\}$$

Thus, this can be done by considering the CDF between  $\left(1 - \frac{\alpha}{2}\right)^{\frac{1}{n}}$  and  $\left(\frac{\alpha}{2}\right)^{\frac{1}{n}}$  interval. But we only want this!

Then,

$$F_U(u) = \int_{-\infty}^u f(u) \, du = \int_{-\infty}^u n u^{n-1} \, du$$

$\left(\frac{\alpha}{2}\right)^{\frac{1}{n}}$

$$\Rightarrow n \int_{(\alpha/2)^{1/n}}^{(1-\alpha/2)^{1/n}} u^{n-1} du = n \left[ \frac{u^{n-1+1}}{n-1+1} \right]_{(\alpha/2)^{1/n}}^{(1-\alpha/2)^{1/n}}$$

$$\Rightarrow n \left[ \frac{u^n}{n} \right]_{(\alpha/2)^{1/n}}^{(1-\alpha/2)^{1/n}} = \left[ (1-\frac{\alpha}{2})^{1/n} \right]^n - \left[ (\frac{\alpha}{2})^{1/n} \right]^n$$

$$\Rightarrow 1 - \frac{\alpha}{2} - \frac{\alpha}{2} = \underline{\underline{1-\alpha}}$$

$$\therefore P\left(\left(\frac{\alpha}{2}\right)^{1/n} < \frac{Y}{\theta} \leq \left(1-\frac{\alpha}{2}\right)^{1/n}\right) = 1-\alpha$$

Hence, verified!

Now, to derive the  $100(1-\alpha)\%$  CI for  $\theta$ ,

Let's begin by considering the equality we verified above.

$$\Rightarrow P\left(\left(\frac{\alpha}{2}\right)^{1/n} < \frac{Y}{\theta} \leq \left(1-\frac{\alpha}{2}\right)^{1/n}\right) = 1-\alpha$$

Dividing the inequality by  $Y$  we get:-

$$P\left(\frac{(\alpha/2)^{1/n}}{Y} < \frac{1}{\theta} \leq \frac{(1-\alpha/2)^{1/n}}{Y}\right) = 1-\alpha$$

Taking reciprocals and redistributing in the inequality above:-

$$P\left(\frac{Y}{(1-\frac{\alpha}{2})^{1/n}} < \theta \leq \frac{Y}{(\alpha/2)^{1/n}}\right) = 1-\alpha$$

∴ The required confidence interval that we get is:-

$$[\because Y = \max(X_i)]$$

$$\Rightarrow \left( \frac{\max(X_i)}{(1 - \frac{\alpha}{2})^{1/n}}, \frac{\max(X_i)}{(\alpha/2)^{1/n}} \right)$$

Ans.

(b)

To verify:-

$$P\left(\alpha^{1/n} \leq \frac{Y}{\theta} \leq 1\right) = 1 - \alpha$$

we consider the LHS first,

$$\Rightarrow P\left(\alpha^{1/n} \leq \frac{Y}{\theta} \leq 1\right)$$

$$\boxed{\therefore \frac{Y}{\theta} = U}$$

$$\Rightarrow P\left(\alpha^{1/n} \leq U \leq 1\right) = F(1) - F(\alpha^{1/n})$$

We can compute by calculating the CDF between this interval which boils down to

$$\Rightarrow \int_{\alpha^{1/n}}^1 f_U(u) du$$

$$\Rightarrow \int_{\alpha^{1/n}}^1 n u^{n-1} du = n \left[ \frac{u^{n-1+1}}{n-1+1} \right] \Big|_{\alpha^{1/n}}^1 = n \left[ \frac{u^n}{n} \right] \Big|_{\alpha^{1/n}}^1$$

$$\Rightarrow 1^n - (\alpha^{1/n})^n \Rightarrow \underline{1 - \alpha}.$$

$\therefore$  We have verified that:-

$$P\left(\alpha^{Y_n} \leq \frac{Y}{\theta} \leq 1\right) = 1-\alpha$$

Ans.

Now, we have to derive  $100(1-\alpha)\% CI$  for  $\theta$ ,

$$P\left(\alpha^{Y_n} \leq \frac{Y}{\theta} \leq 1\right) = 1-\alpha$$

Dividing the inequality by  $Y$ ,

$$P\left(\frac{\alpha^{Y_n}}{Y} \leq \frac{1}{\theta} \leq \frac{1}{Y}\right) = 1-\alpha$$

Taking reciprocals of the inequality above, we get:-

$$P\left(Y \leq \theta \leq \frac{Y}{\alpha^{Y_n}}\right) = 1-\alpha$$

$$\therefore Y = \max(X_i),$$

$$P\left(\max(X_i) \leq \theta \leq \frac{\max(X_i)}{\alpha^{Y_n}}\right) = 1-\alpha$$

$\therefore$  The required confidence interval is

$$\left(\max(X_i), \frac{\max(X_i)}{\alpha^{Y_n}}\right)$$

Ans

(c) Confidence Interval for part (a) is :-

$$\left( \frac{\max(x_i)}{(1-\alpha/2)^{1/n}}, \frac{\max(x_i)}{(\alpha/2)^{1/n}} \right) \quad \text{--- (1)}$$

~~Confidence~~ Confidence Interval for part (b) is :-

$$\left( \max(x_i), \frac{\max(x_i)}{\alpha^{1/n}} \right) \quad \text{--- (2)}$$

for finding which of these intervals is shorter, for which we need to find the length of the intervals first which is, (Upper bound - lower bound).

Let  $L_1$  be the length of CI from (1),

$$L_1 = \frac{\max(x_i)}{(\alpha/2)^{1/n}} - \frac{\max(x_i)}{(1-\alpha/2)^{1/n}} = \max(x_i) \left[ \frac{1}{(\alpha/2)^{1/n}} - \frac{1}{(1-\alpha/2)^{1/n}} \right]$$

$$= \max(x_i) \cdot 2^{1/n} \cdot \left[ \frac{1}{\alpha^{1/n}} - \frac{1}{(2-\alpha)^{1/n}} \right]$$

$$\therefore L_1 = 2^{1/n} \cdot \max(x_i) \cdot \left[ \frac{(2-\alpha)^{1/n} - \alpha^{1/n}}{\alpha^{1/n}(2-\alpha)^{1/n}} \right] \quad \text{--- (3)}$$

Let  $L_2$  be the length of CI from (2),

$$L_2 = \frac{\max(x_i)}{\alpha^{1/n}} - \min(x_i) = \max(x_i) \left[ \frac{1}{\alpha^{1/n}} - 1 \right]$$

$$\therefore L_2 = \max(x_i) \left[ \frac{1 - \alpha^{1/n}}{\alpha^{1/n}} \right] \quad \text{--- (4)}$$

Multiplying (3) and (4) by  $\alpha^{1/n} \cdot (2-\alpha)^{1/n}$ , we get:-

$$L_1 (\alpha^{\frac{1}{n}}) (2-\alpha)^{\frac{1}{n}} = 2^{\frac{1}{n}} \cdot \max(x_i) \alpha^{\frac{1}{n}} (2-\alpha)^{\frac{1}{n}}$$

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

$$\Rightarrow L_1 \alpha^{\frac{1}{n}} (2-\alpha)^{\frac{1}{n}} = 2^{\frac{1}{n}} \cdot \max(x_i) \left[ (2-\alpha)^{\frac{1}{n}} - \alpha^{\frac{1}{n}} \right] \quad (5)$$

And for eq (4),

$$L_2 (\alpha^{\frac{1}{n}}) (2-\alpha)^{\frac{1}{n}} = \max(x_i) \alpha^{\frac{1}{n}} (2-\alpha)^{\frac{1}{n}} \cdot \left[ \frac{1 - \alpha^{\frac{1}{n}}}{\alpha^{\frac{1}{n}}} \right]$$

$$\Rightarrow L_2 \alpha^{\frac{1}{n}} (2-\alpha)^{\frac{1}{n}} = (2-\alpha)^{\frac{1}{n}} \max(x_i) \cdot [1 - \alpha^{\frac{1}{n}}] \quad (6)$$

Now, to determine a relationship between (5)th and (6)th, we need to find a definite relationship among their individual terms, we might need to use the fact that,

$$0 < \alpha < 1 \quad (7)$$

Multiplying inequality (7) by  $-1$ ,

$$0 > -\alpha > -1$$

Add 2 to both sides of the inequality (full inequality),

$$2 > 2 - \alpha > 2 - 1$$

$$\Rightarrow \underline{2 > 2 - \alpha > 1}$$

Now, we will take the power  $\frac{1}{n}$  on the full inequality,

$$2^{\frac{1}{n}} > (2-\alpha)^{\frac{1}{n}} > 1$$

$$\therefore 1 < (2-\alpha)^{\frac{1}{n}} < 2^{\frac{1}{n}} \quad (8)$$

From eq (5),

$$L_1 \alpha^{\frac{1}{n}} (2-\alpha)^{\frac{1}{n}} = 2^{\frac{1}{n}} \max(x_i) \cdot [(2-\alpha)^{\frac{1}{n}} - \alpha^{\frac{1}{n}}]$$

If we use the latter half of the inequality ⑧ in the above equation, it becomes:-

$$L_1 \alpha^{1/n} \cdot (2-\alpha)^{1/n} > \max(x_i) \cdot (2-\alpha)^{1/n} [ (2-\alpha)^{1/n} - \alpha^{1/n} ]$$

$\left[ \because (2-\alpha)^{1/n} < 2^{1/n} \right]$

Now, from inequality ⑧,  $\therefore 1 < (2-\alpha)^{1/n}$ ,

$$L_1 \alpha^{1/n} \cdot (2-\alpha)^{1/n} > (2-\alpha)^{1/n} \cdot \max(x_i) [ 1 - \alpha^{1/n} ]$$

From eqn ⑥, we know that RHS of the above inequality becomes,

$$L_1 \alpha^{1/n} \cdot (2-\alpha)^{1/n} > L_2 \alpha^{1/n} (2-\alpha)^{1/n}$$

Dividing both sides of the above inequality by  $\alpha^{1/n} (2-\alpha)^{1/n}$  and since both are positive terms, the sign doesn't change:-

$$\boxed{L_1 > L_2}$$

$\therefore$  we have found that  $L_2$  has the narrower length between  $L_1$  and  $L_2$ .

The narrower length is  $\boxed{L_2 = \max(x_i) \left[ \frac{1 - \alpha^{1/n}}{\alpha^{1/n}} \right]}$

Ans.

Using the data provided in the problem as follows:-

$$x_1 = 4.2, x_2 = 3.5, x_3 = 1.7, x_4 = 1.2, x_5 = 2.5$$

and  $\gamma = 0.95$  and  $n = 5$ .

Now, CI with narrower length is  $\Rightarrow$

$$CI_2 = \left( \max(x_i), \frac{\max(x_i)}{\alpha^{1/n}} \right) \quad \text{--- ②}$$

$$\therefore 1 - \alpha = \gamma \Rightarrow 1 - \alpha = 0.95 \Rightarrow \boxed{\alpha = 0.05}$$

And ,

$$\max(x_i) = \max(4.2, 3.5, 1.7, 1.2, 2.4) \\ = 4.2$$

Substituting these values in CI-② ,

$$CI = \left( 4.2, \frac{4.2}{(0.05)^{1/5}} \right) \quad (\because n=5)$$

$$\Rightarrow CI = (4.2, 7.65) \quad \underline{\text{Ans}}$$

$\therefore$  95% CI for  $\theta$  with above values is  $\Rightarrow$   
 $\underline{(4.2, 7.65)}$  Ans.

4.

Let  $X_1, \dots, X_n$  be iid  $N(\mu, \sigma^2)$  - distributed.

Also,  $\mu$  and  $\sigma^2$  are unknown parameters here.

Now, to construct the CI for this:-

Step ①:- find a pivot for unknown parameter and to calculate the CI for  $\sigma^2$ , we get the pivot:-

$$\boxed{\Xi = \frac{(n-1)S_n^2}{\sigma^2}}$$

(As shown in the class).

Step ②:- Determining the distribution of the pivot,

As shown in the class, the distribution of  $\Xi$  is basically  $\chi^2_{n-1}$  distribution. (with  $n-1$  degrees of freedom).

Now, let  $q_\gamma$  be the quantile function.

$q_\gamma(x) =$  The value of  $x$  such that,

$$\boxed{P(\chi^2_{n-1} \leq x) = \gamma}$$

Step ③:-

As we need to construct a confidence interval, we need to choose  $X_1$  and  $X_2$ .

And,

$$\boxed{X_1 + X_2 = 1-\gamma}$$

holds where  $\gamma$  is the confidence level.

$$\text{Now, } P(q_\gamma(X_1) \leq \Xi(X_1, \dots, X_n; \mu, \sigma^2) \leq q_\gamma(\chi^2_{n-1}, 1-\gamma)) = \gamma$$

$$P(q_\gamma(X_1) \leq \Xi(X_1, \dots, X_n; \mu, \sigma^2) \leq q_\gamma(\chi^2_{n-1}, 1-\gamma)) = \gamma$$

Substituting the pivot in the equation above,

$$P\left(q_V(\eta_{n-1}^2, \alpha_1) \leq \frac{(n-1)s_n^2}{\sigma^2} \leq q_V(\eta_{n-1}^2, 1-\alpha_2)\right) = \gamma$$

Step(4):- Now, solving for  $\sigma^2$  as we need to create CI for  $\sigma^2$ .

$$P\left(q_V(\eta_{n-1}^2, \alpha_1) \leq \frac{(n-1)s_n^2}{\sigma^2} \leq q_V(\eta_{n-1}^2, 1-\alpha_2)\right) = \gamma$$

Dividing the inequality by  $(n-1)s_n^2$ , we get:-

$$P\left(\frac{q_V(\eta_{n-1}^2, \alpha_1)}{(n-1)s_n^2} \leq \frac{1}{\sigma^2} \leq \frac{q_V(\eta_{n-1}^2, 1-\alpha_2)}{(n-1)s_n^2}\right) = \gamma$$

Now, we take reciprocals of the terms in the inequality through cross-products  $\Rightarrow$

$$P\left(\frac{(n-1)s_n^2}{q_V(\eta_{n-1}^2, 1-\alpha_2)} \leq \sigma^2 \leq \frac{(n-1)s_n^2}{q_V(\eta_{n-1}^2, \alpha_1)}\right) = \gamma$$

Therefore,

The confidence interval is  $\Rightarrow$

$$\left[\frac{(n-1)s_n^2}{q_V(\eta_{n-1}^2, 1-\alpha_2)}, \frac{(n-1)s_n^2}{q_V(\eta_{n-1}^2, \alpha_1)}\right] \quad \underline{\text{Ans}}$$

Now, for the calculation part,

$$\gamma = 0.99, \quad n = 15 \quad \text{and} \quad S_n^2 = 20$$

We need a CI of the form  $(0, R_n)$ .

$$\therefore \alpha_1 + \alpha_2 = 1 - \gamma \Rightarrow \alpha_1 + \alpha_2 = 1 - 0.99 = 0.01$$

for the  $(0, R_n)$  interval, the lower bound of the CI found should be zero,

$$\therefore \frac{(n-1) S_n^2}{\sqrt{\gamma(\gamma_{n-1}^2, 1-\alpha_2)}} = 0$$

$\therefore S_n$  and  $n$ , both are finite and  $n \neq 1$ ,

$$\Rightarrow (\therefore \sqrt{\gamma(\gamma_{n-1}^L, 1-\alpha_2)} = \infty)$$

~~$\therefore$~~  we know that:  $\sqrt{\gamma(\pm)} = \infty$ ,

$$\therefore 1 - \alpha_2 = 1 \Rightarrow \boxed{\alpha_2 = 0}$$

$$\therefore \alpha_1 + 0 = 0.01 \Rightarrow \boxed{\alpha_1 = 0.01}$$

$$\therefore R_n \Rightarrow \frac{(n-1) S_n^2}{\sqrt{\gamma(\gamma_{n-1}^2, 0.01)}}$$

$$CI \Rightarrow \left( 0, \frac{(n-1) S_n^2}{\sqrt{\gamma(\gamma_{n-1}^2, 0.01)}} \right)$$

Putting in the values,

$(n-1)$  degrees of freedom  $\Rightarrow (15-1) = 14$  degrees of freedom.

$$\therefore CI \Rightarrow \left( 0, \frac{14 \times (20)^2}{\sqrt{\chi^2_{15-1}, 0.01}} \right)$$

$$\Rightarrow \left( 0, \frac{14 \times 20^2}{4.66} \right) \quad \left[ \because \sqrt{\chi^2_{14, 0.01}} = 4.66 \right]$$

$$\Rightarrow \left( 0, \frac{5600}{4.66} \right)$$

$$\therefore CI \Rightarrow (0, 1201.7167)$$

$$\Rightarrow \boxed{(0, 1201.72)} \quad \underline{\text{Ans}}$$

5.

$$\therefore n = 50, \bar{X}_n = 634.16 \text{ and } S_n = 164.43$$

Here  $\sigma^2$  and  $\mu$  are unknown variables.

Now, we will try to construct the confidence interval for  $\mu \Rightarrow$ .

Step ①:- find a pivot for the unknown parameters,

$$I(X_1, \dots, X_n; \mu, \sigma^2) = \frac{\bar{X}_n - \mu}{S_n / \sqrt{n}} \rightarrow \begin{array}{l} \text{This will be} \\ \text{the approximate} \\ \text{pivot when considering} \\ n \text{ as large.} \end{array}$$

Step ②:- finding the distribution of the pivot,

The distribution of  $I$  is approximately  $N(0, 1)$  [as shown in class]

$$I = \frac{\bar{X}_n - \mu}{S_n / \sqrt{n}} \stackrel{\text{approx}}{\sim} N(0, 1)$$

Let  $q$  be the quantile function,

$q(\alpha) =$  The value of  $x$  such that,

$$P(N(0, 1) \leq x) = \alpha$$

Step ③:- As we need to construct a confidence interval, we need to choose  $\alpha_1$  &  $\alpha_2$ , such that,

$$\alpha_1 + \alpha_2 = 1 - \gamma$$

Now,

$$P(q_{\gamma}(N(0,1), \alpha_1) \leq \bar{X}_n - \mu \leq q_{\gamma}(N(0,1), 1-\alpha_2)) = \gamma$$

$\Rightarrow$

$$P(q_{\gamma}(N(0,1), \alpha_1) \leq \frac{\bar{X}_n - \mu}{S_n / \sqrt{n}} \leq q_{\gamma}(N(0,1), 1-\alpha_2)) = \gamma \quad \boxed{①}$$

Step(4): Solving for  $\mu$ ,

Multiplying the inequality in eqn ① above by  $S_n / \sqrt{n}$ ,

$$P\left(\frac{S_n}{\sqrt{n}} q_{\gamma}(N(0,1), \alpha_1) \leq \bar{X}_n - \mu \leq \frac{S_n}{\sqrt{n}} q_{\gamma}(N(0,1), 1-\alpha_2)\right) = \gamma$$

Subtracting  $\bar{X}_n$  from all sides of the inequality above,

$$P\left(\frac{S_n}{\sqrt{n}} q_{\gamma}(N(0,1), \alpha_1) - \bar{X}_n \leq -\mu \leq \frac{S_n}{\sqrt{n}} q_{\gamma}(N(0,1), 1-\alpha_2) - \bar{X}_n\right) = \gamma$$

Multiplying all the sides by  $-1$ , we get:-

$$P\left(\bar{X}_n - \frac{S_n}{\sqrt{n}} q_{\gamma}(N(0,1), \alpha_1) \leq \mu \leq \bar{X}_n - \frac{S_n}{\sqrt{n}} q_{\gamma}(N(0,1), 1-\alpha_2)\right) = \gamma$$

$\underbrace{\hspace{30em}}$   $\underbrace{\hspace{30em}}$   $\boxed{②}$

$\therefore \gamma = 0.95$  (95%) and  $\alpha_1 + \alpha_2 = 1 - \gamma$ ,

$$\text{we get } \Rightarrow \alpha_1 + \alpha_2 = 1 - 0.95 \Rightarrow \boxed{\alpha_1 + \alpha_2 = 0.05}$$

We consider this to be a balanced CI,

$$(\alpha_1 = \alpha_2) \Rightarrow 2\alpha_1 + \alpha_1 = 0.05$$

$$\Rightarrow 2\alpha_1 = 0.05 \Rightarrow \boxed{\alpha_1 = 0.025}$$

$$\boxed{\alpha_1 = \alpha_2 = 0.025}$$

Putting all the values in the inequality ② above,

$$654.16 - \frac{164.43}{\sqrt{50}} q_{(N(0,1), 0.975)} \leq \mu \leq 654.16 - \frac{164.43}{\sqrt{50}} q_{(N(0,1), 0.025)}$$

$$\therefore q_{(N(0,1), 0.975)} = \underline{\underline{-1.96}}$$

$$\text{and } q_{(N(0,1), 0.975)} = -q_{(N(0,1), 0.025)}$$

$$\therefore q_{(N(0,1), 0.025)} = \underline{\underline{-1.96}}$$

Therefore,

$$654.16 - \frac{164.43}{\sqrt{50}} \times (1.96) \leq \mu \leq 654.16 - \frac{164.43}{\sqrt{50}} \times (-1.96)$$

$\Rightarrow$

$$\boxed{608.58 \leq \mu \leq 699.74}$$

Aho.



b.)  $\sigma^2 = 175$ ,  $L = 50 \text{ ppm}$ ,  $\gamma = 0.95$

for constructing the CI for this:-

Step ①:- find a pivot for the unknown parameters,

$$\mathbb{E}(x_1, \dots, x_n; \mu, \sigma^2) = \frac{\bar{x}_n - \mu}{\sigma \sqrt{n}} \quad (\text{Approx. pivot}),$$

Step ②:- find the distribution of the pivot,

The distribution for  $\mathbb{E}$  is  $N(0, 1)$ .

$$\boxed{\mathbb{E} = \frac{\bar{x}_n - \mu}{\sigma \sqrt{n}} \sim N(0, 1)}$$

Let  $q_\gamma$  be its quantile function,

$q_\gamma(\alpha) = \text{The value of } x \text{ such that:-}$

$$P(N(0, 1) \leq x) = \alpha$$

Step ③:- Choosing  $\alpha_1$  and  $\alpha_2$  as we need to construct bounds for CI such that:-

$$\boxed{\alpha_1 + \alpha_2 = 1 - \gamma}$$

$$P(q_\gamma(\alpha_1) \leq \mathbb{E}(x_1, \dots, x_n; \mu) \leq q_\gamma(1 - \alpha_2)) = \gamma$$

Substituting the value of the pivot,

$$P(q_\gamma(N(0, 1), \alpha_1) \leq \frac{\bar{x}_n - \mu}{\sigma \sqrt{n}} \leq q_\gamma(N(0, 1), 1 - \alpha_2)) = \gamma$$

Step ④:- Solving the inequality for  $\mu$ ,

$$P\left(\frac{\sigma}{\sqrt{n}} q_{\gamma}(N(0,1), \alpha_1) \leq \bar{X}_n - \mu \leq \frac{\sigma}{\sqrt{n}} q_{\gamma}(N(0,1), 1-\alpha_2)\right) = \gamma$$

$$P\left(\frac{\sigma}{\sqrt{n}} q_{\gamma}(N(0,1), \alpha_1) - \bar{X}_n \leq -\mu \leq \frac{\sigma}{\sqrt{n}} q_{\gamma}(N(0,1), 1-\alpha_2) - \bar{X}_n\right) = \gamma$$

$$\Rightarrow P\left(\bar{X}_n - \frac{\sigma}{\sqrt{n}} q_{\gamma}(N(0,1), 1-\alpha_2) \leq \mu \leq \bar{X}_n - \frac{\sigma}{\sqrt{n}} q_{\gamma}(N(0,1), \alpha_1)\right) = \gamma$$

$$\therefore CI \Rightarrow \left(\bar{X}_n - \frac{\sigma}{\sqrt{n}} q_{\gamma}(N(0,1), 1-\alpha_2), \bar{X}_n - \frac{\sigma}{\sqrt{n}} q_{\gamma}(N(0,1), \alpha_1)\right)$$

$\therefore$  Length of CI ( $L$ )  $\Rightarrow$

$$= \left[ \bar{X}_n - \frac{\sigma}{\sqrt{n}} q_{\gamma}(N(0,1), \alpha_1) - \left( \bar{X}_n - \frac{\sigma}{\sqrt{n}} q_{\gamma}(N(0,1), 1-\alpha_2) \right) \right]$$

$$= \frac{\sigma}{\sqrt{n}} q_{\gamma}(N(0,1), 1-\alpha_2) - \frac{\sigma}{\sqrt{n}} q_{\gamma}(N(0,1), \alpha_1)$$

$$L = \frac{\sigma}{\sqrt{n}} [q_{\gamma}(N(0,1), 1-\alpha_2) - q_{\gamma}(N(0,1), \alpha_1)]$$

$$\therefore q_{\gamma}(N(0,1), 1-\alpha_2) = -q_{\gamma}(N(0,1), \alpha_1)$$

$$\therefore L = \frac{2\sigma}{\sqrt{n}} q_{\gamma}(N(0,1), \frac{\alpha_1}{1-\alpha_2})$$

$$\therefore \alpha_1 + \alpha_2 = 1 - \gamma \Rightarrow \alpha_1 + \alpha_2 = 1 - 0.95 = 0.05$$

$$\therefore \text{Balanced CI } (\alpha_1 = \alpha_2) \Rightarrow \alpha_1 = \alpha_2 = \frac{0.05}{2} = 0.025$$

for  $n$ ,

$$n = \left( \frac{2\sigma}{L} q_{\mathcal{N}(0,1), 1-\alpha_1} \right)^2$$

$$n = \left( \frac{2 \times 17.5}{50} q_{\mathcal{N}(0,1), 1-0.025} \right)^2$$

$\Rightarrow$

$$n = \left( \frac{350}{50} \times 1.96 \right)^2 \quad [ \because q_{\mathcal{N}(0,1), 0.975} = 1.96 ]$$

$$n = 188.2384$$

$$\Rightarrow \boxed{n \approx 189} \quad (\text{Rounded up})$$

$\therefore$  Required sample size  $\Rightarrow \underline{189}$

Ans.

(6.) Let  $x_1, \dots, x_n$  form a random sample from the exponential distribution with unknown parameter  $\lambda$ .

We know that the pdf  $f(x)$  for the exponential distribution is given by:-

$$f(x) = \begin{cases} \lambda e^{-\lambda x} & , x > 0 \\ 0 & , \text{otherwise} \end{cases}$$

Here,  $\lambda$  is the unknown parameter.

Now, let  $y = 2\lambda x$ ,

the pdf  $g(y)$  of the exponential distribution can now be represented as:-

$$g(y) = \begin{cases} \frac{1}{2} e^{-y/2} & , y > 0 \\ 0 & , \text{otherwise} \end{cases} \quad \text{--- (1)}$$

Also, we know that,

$$\begin{aligned} \text{Gamma } (\frac{1}{2}, \lambda) &= \text{Exp}(\lambda) \quad \text{--- (2)} \\ \text{and, } \chi_m^2 &= \text{Gamma } (\frac{m}{2}, \frac{1}{2}) \quad \text{--- (3)} \end{aligned} \quad \left. \begin{array}{l} \text{from the} \\ \text{cheat sheet} \\ \text{provided.} \end{array} \right\}$$

Putting  $m=2$  in equation (3), we get:-

$$\chi_2^2 = \text{Gamma } (\frac{1}{2}, \frac{1}{2})$$

Using equation (2),

$$\chi_2^2 = \text{Gamma } (\frac{1}{2}, \frac{1}{2}) = \text{Exp}(\frac{1}{2}) \quad \text{--- (4)}$$

There are huge similarities between the pdf  $f(x)$  and pdf  $g(y)$ . It is clear that  $g(y)$  follows an exponential distribution but with parameter  $\frac{1}{2}$ , since

$$f(x) = \lambda e^{-\lambda x} \quad x > 0 \quad [0, \text{ otherwise}]$$

$$g(y) = \frac{1}{2} e^{-y/2} \quad y > 0 \quad [0, \text{ otherwise}]$$

from equation (4), we can come to the conclusion that  $f(y)$  has a chi-squared distribution with degree of freedom 2.

Now, let's try to construct the required confidence intervals.

Step ①:— find an appropriate pivot for the unknown parameter,

$$\begin{aligned} \mathbb{E}(x_1, \dots, x_n; \lambda) &= 2\lambda \sum_{i=1}^n x_i \\ &= \sum_{i=1}^n y_i \quad [\because y = 2\lambda x \text{ as assumed before}] \end{aligned}$$

Step ②:— find the distribution of the pivot,

As seen above,

$$\mathbb{E}(x_1, \dots, x_n; \lambda) = 2\lambda \sum_{i=1}^n x_i = \sum_{i=1}^n y_i$$

As discussed before,  $y_i = 2\lambda x_i$  and it follows an exponential distribution with parameter  $1/2$ . This is equivalent to chi-squared distribution with 2 degrees of freedom.

Further all  $y_i$ 's are independent,  $\therefore$ , the pivot  $\mathbb{E}$  follows a chi-squared distribution with  $2n$  degrees of freedom.

$$\mathbb{E}(X_1, \dots, X_n; \lambda) \sim \chi_{2n}^2$$

Let  $q_\gamma$  be its quantile function such that:-

$q_\gamma(\alpha) =$  the value of  $x$  such that:-

$$P(\chi_{2n}^2 \leq x) = \alpha$$

Step 3:- Choose values of  $\alpha_1$  and  $\alpha_2$  such that:-

$$\boxed{\alpha_1 + \alpha_2 = 1 - \gamma}$$

$$P(q_\gamma(\chi_{2n}^2, \alpha_1) \leq \mathbb{E}(X_1, \dots, X_n; \lambda) \leq q_\gamma(\chi_{2n}^2, 1 - \alpha_2)) = \gamma$$

Substituting the value of the pivot in the equation above:-

$$P(q_\gamma(\chi_{2n}^2, \alpha_1) \leq 2 \lambda \sum_{i=1}^n X_i \leq q_\gamma(\chi_{2n}^2, 1 - \alpha_2)) = \gamma$$

Step 4:- Solve the inequality for  $\lambda$ , we get:-

Dividing both sides by  $2 \sum_{i=1}^n X_i$ , we get:-

$$P\left(\frac{1}{2 \sum_{i=1}^n X_i} q_\gamma(\chi_{2n}^2, \alpha_1) \leq \lambda \leq \frac{1}{2 \sum_{i=1}^n X_i} q_\gamma(\chi_{2n}^2, 1 - \alpha_2)\right) = \gamma$$

$$\therefore \text{The CI is } \Rightarrow \left( \frac{q_\gamma(\chi_{2n}^2, \alpha_1)}{2 \sum_{i=1}^n X_i}, \frac{q_\gamma(\chi_{2n}^2, 1 - \alpha_2)}{2 \sum_{i=1}^n X_i} \right)$$

An.

7

Sample mean ( $\bar{X}_n$ ) = 3.2 $n=100$  (100 iid poisson variables).Hence the unknown parameter is  $\theta$ .

Step(1)

Since, we want to find an approximate confidence interval, we can consider the pivot to be  $\Rightarrow$

$$\boxed{\Xi(x_1, \dots, x_n; \mu, \theta) \Rightarrow \frac{\bar{X}_n - \mu}{\theta/\sqrt{n}}}$$

Step(2) - finding the distribution of the pivot,

$\therefore$  we have taken the above pivot as an approximate considering that  $n$  is large,

$$\boxed{\Xi = \frac{\bar{X}_n - \mu}{\theta/\sqrt{n}} \sim N(0, 1)}$$

Let  $q(\alpha)$  be the quantile function such that :-

$$\underline{P(N(0,1) \leq x) = \alpha}$$

Step(3) - But first we need to find  $\mu$  and  $\theta$  for the poisson distribution, ( $\because$  this is a discrete function)

$$\therefore E[X] = \mu \Rightarrow \boxed{E[X] = \sum_{x \geq 0} x f(x)}$$

for Poisson distribution, we know that:

$$\boxed{f(x) = \frac{e^{-\theta} \theta^x}{x!}}$$

$$\therefore E[x] = \sum_{n \geq 0} n \cdot \frac{e^{-\theta} \cdot \theta^n}{n!} = \sum_{n \geq 0} \frac{n \cdot e^{-\theta} \cdot \theta^n}{n \cdot ((n-1)!)}$$

$$\Rightarrow E[x] = \sum_{n \geq 1} \frac{e^{-\theta} \cdot \theta^n}{(n-1)!} = e^{-\theta} \cdot \theta \sum_{n \geq 1} \frac{\theta^{n-1}}{(n-1)!}$$

By Taylor series expansion,

$$\boxed{\sum_{n \geq 1} \frac{\theta^{n-1}}{(n-1)!} = e^\theta}$$

$$\therefore E[x] = e^{-\theta} \cdot \theta \cdot e^\theta = \theta$$

$$\boxed{\therefore E[x] = \theta}$$

Now, we need to compute  $V[x]$  (Variance),

$$\boxed{V[x] = E[x^2] - (E[x])^2} \quad \textcircled{1}$$

$\because$  we know  $E[x]$ , let's compute  $E[x^2]$ .

$$E[x^2] = \sum_{n \geq 0} n^2 f(n) = \sum_{n \geq 0} n^2 \frac{e^{-\theta} \cdot \theta^n}{n!}$$

$$= \sum_{n \geq 0} \frac{n \cdot n \cdot e^{-\theta} \cdot \theta^{n-1} \cdot \theta}{n!((n-1)!)}$$

$$\Rightarrow E[x^2] = e^{-\theta} \cdot \theta \sum_{n \geq 1} \frac{(n-1+1) \theta^{n-1}}{(n-1)!} = e^{-\theta} \cdot \theta \left[ \sum_{n \geq 1} \frac{(n-1) \theta^{n-1}}{(n-1)!} + \sum_{n \geq 1} \frac{1 \cdot \theta^{n-1}}{(n-1)!} \right]$$

By Taylor series expansion, we know that:

$$\sum_{n \geq 1} \frac{\theta^{n-1}}{(n-1)!} = e^\theta$$

$$\therefore \Rightarrow E[X^2] = e^{-\theta} \cdot \theta \left[ \sum_{n \geq 1} \frac{(n-1) \cdot \theta^{n-2} \cdot \theta}{(n-1)(n-2)!} + e^\theta \right]$$

$$\Rightarrow E[X^2] = e^{-\theta} \cdot \theta \underbrace{\left[ \theta \sum_{n \geq 2} \frac{\theta^{n-2}}{(n-2)!} + e^\theta \right]}_{\downarrow e^\theta}$$

$$\Rightarrow E[X^2] = e^{-\theta} \cdot \theta \left[ \theta e^\theta + e^\theta \right] = e^{-\theta} \cdot \theta \cdot e^\theta [\theta + 1] \\ = \theta(\theta + 1)$$

Putting values in ①,

$$V[X] = \theta(\theta + 1) - (\theta)^2 = \theta^2 + \theta - \theta^2 = \theta$$

$$\therefore V[X] = \theta$$

Also, we will choose  $\alpha_1$  and  $\alpha_2$  as we need to construct bounds for C2 such that:-

$$\boxed{\alpha_1 + \alpha_2 = 1 - \gamma}$$

$$P(g(N(\alpha_1), \alpha_1) \leq \bar{x}(x_1, \dots, x_n; \mu, \sigma)) \leq g(N(\alpha_1), 1 - \alpha_2) \\ = \gamma$$

Substituting the value of the pivot,

$$P(q_{\gamma}(N(0,1), \alpha_1) \leq \frac{\bar{X}_n - \mu}{\sigma/\sqrt{n}} \leq q_{\gamma}(N(0,1), 1-\alpha_2)) = \gamma$$

Step (4):-

Now, we can put in the values for  $\mu$  and  $\sigma$  here. (And start solving for  $\theta$ ).

~~Always by the law of large numbers (CLT), we know that sample mean becomes the actual mean population mean when n is large, which is the case here.~~

As computed before,  $E[x] = \mu = \theta$  and  $V[x] = \sigma^2 = \theta$ ,

Thus,

$$P(q_{\gamma}(N(0,1), \alpha_1) \leq \frac{\bar{X}_n - \theta}{\sqrt{\theta}/\sqrt{n}} \leq q_{\gamma}(N(0,1), 1-\alpha_2)) = \gamma$$

Multiplying inequality by  $\sqrt{\theta/n}$ ,

$$P\left(\sqrt{\frac{\theta}{n}} q_{\gamma}(N(0,1), \alpha_1) \leq \bar{X}_n - \theta \leq \sqrt{\frac{\theta}{n}} q_{\gamma}(N(0,1), 1-\alpha_2)\right) = \gamma$$

Subtracting  $\bar{X}_n$  from the inequality,

$$P\left(\sqrt{\frac{\theta}{n}} q_{\gamma}(N(0,1), \alpha_1) - \bar{X}_n \leq -\theta \leq \sqrt{\frac{\theta}{n}} q_{\gamma}(N(0,1), 1-\alpha_2) - \bar{X}_n\right) = \gamma$$

Multiplying inequality by  $-1$ ,

$$P\left(\bar{X}_n - \sqrt{\frac{\theta}{n}} q_{\gamma}(N(0,1), 1-\alpha_2) \leq \theta \leq \bar{X}_n - \sqrt{\frac{\theta}{n}} q_{\gamma}(N(0,1), \alpha_1)\right) = \gamma$$

Now, by the law of large numbers (CLT), we know that sample mean becomes the population mean and  $\therefore$ ,

$$E[x] = V[x] = \theta, \therefore \underline{\bar{X}_n},$$

So,  
it becomes,

$$P \left( \bar{X}_n - \sqrt{\frac{\bar{X}_n}{n}} q(N(0,1), 1-\alpha_2) \leq \theta \leq \bar{X}_n - \sqrt{\frac{\bar{X}_n}{n}} q(N(0,1), \alpha_1) \right) = \gamma$$

Also,

$$\therefore \alpha_1 + \alpha_2 = 1 - \gamma \quad \text{and} \quad \gamma = 0.95$$

$$\Rightarrow \alpha_1 + \alpha_2 = 1 - 0.95 = 0.05$$

for a balanced CI,

$$\boxed{\alpha_1 = \alpha_2}$$

$$\therefore 2\alpha_1 = 0.05 \Rightarrow \boxed{\alpha_1 = 0.025}$$

$$\Rightarrow \boxed{\alpha_1 = \alpha_2 = 0.025}$$

To calculate 95% CI, we substitute all the value for the confidence interval,

$$CI \Rightarrow \left( \bar{X}_n - \sqrt{\frac{\bar{X}_n}{n}} q(N(0,1), 1-\alpha_2) \leq \theta \leq \bar{X}_n - \sqrt{\frac{\bar{X}_n}{n}} q(N(0,1), \alpha_1) \right)$$

$$CI \Rightarrow \left( 3.2 - \sqrt{\frac{3.2}{100}} q(N(0,1), 1-0.025) \leq \theta \leq 3.2 - \sqrt{\frac{3.2}{100}} q(N(0,1), 0.025) \right)$$

$$\therefore q_{\sqrt{N(0,1)}}(1-0.025) = q_{\sqrt{N(0,1)}}(0.975) = \underline{1.96}$$

and

$$q_{\sqrt{N(0,1)}}(0.025) = \underline{-1.96}$$

∴

$$\Rightarrow CI = \left( 3.2 - 1.96 \times \sqrt{\frac{3.2}{100}} \leq \theta \leq 3.2 + 1.96 \sqrt{\frac{3.2}{100}} \right)$$

$$\Rightarrow \boxed{2.8494 \leq \theta \leq 3.5506}$$

∴ 95%-CI for  $\theta$  of the poisson distribution

$$\text{is } \Rightarrow (2.8494, 3.5506)$$

$$= \underline{(2.85, 3.55)}$$

Ans.