

concept of large sample confidence interval
If σ^2 is unknown and $n > 30$
then $100(1-\alpha)\%$ C.I. for μ can be given by
$$\left[\bar{x} - z_{\alpha/2} \frac{s}{\sqrt{n}}, \bar{x} + z_{\alpha/2} \frac{s}{\sqrt{n}} \right]$$

ch-9

(2)

$\sigma = 40$
 $n = 30, \bar{x} = 780$

$100(1-\alpha)\% = 96\%$

$1-\alpha = 0.96$
 $\alpha = 0.04$

$z_{\alpha/2} = z_{0.02}$

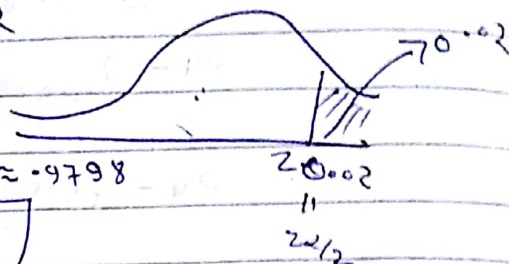
Since $P(Z > z_{0.02}) = 0.02$

$1 - P(Z > z_{0.02}) = 0.98$

$P(Z < z_{0.02}) = 1 - 0.02$

$P(Z < z_{0.02}) = 0.98 \approx 0.9798$

$z_{0.02} = 2.05$



confidence interval for μ

$$\bar{x} - z_{\alpha/2} \frac{\sigma}{\sqrt{n}} < \mu < \bar{x} + z_{\alpha/2} \frac{\sigma}{\sqrt{n}}$$

$$780 - \frac{(2.05) \cdot 40}{\sqrt{30}} < \mu < 780 + \frac{(2.05) \cdot 40}{\sqrt{30}}$$

$$765.02 < \mu < 794.97$$

(4)

$n = 50, \bar{x} = 174.5, s = 6.9$

(a) $100(1-\alpha)\% = 98\%$

$1-\alpha = 0.98$

$\alpha = 0.02, \alpha/2 = 0.01$

$z_{\alpha/2} = z_{0.01}$

Since $P(Z > z_{\alpha/2}) = \alpha/2$

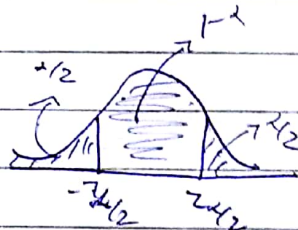
$P(Z > z_{0.01}) = 0.01$

$P(Z > z_{0.01}) = 0.01$

$P(Z < z_{0.01}) = 1 - 0.01 = 0.99$

$P(Z < z_{0.01}) = 0.99 \approx 0.9901$

$z_{0.01} = 2.33$



Since confidence interval for μ

$$\bar{x} - z_{\alpha/2} \frac{s}{\sqrt{n}} < \mu < \bar{x} + z_{\alpha/2} \frac{s}{\sqrt{n}}$$

$$174.5 - \frac{2.33 \times 6.9}{\sqrt{50}} < \mu < 174.5 + \frac{2.33 \times 6.9}{\sqrt{50}}$$

$$173.226 < \mu < 176.77$$

$$\Rightarrow 173.23 < \mu < 176.77$$

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(b) (Th: If \bar{x} is used as an estimate of μ , we can be $100(1-\alpha)\%$ confident that error will not exceed $z_{\alpha/2} \frac{s}{\sqrt{n}}$.)

$$\text{ie. } e < z_{\alpha/2} \frac{s}{\sqrt{n}}$$

$$z_{\alpha/2} = 2.33, s = 6.9$$

$$e < 2.33 \cdot \frac{6.9}{\sqrt{50}}$$

$$, n = 50$$

$$e < 2.273$$

$$\Rightarrow |e| < 2.27$$

(5)

$$n = 100, \bar{x} = 23,500, s = 3900$$

✓

(a)

$$100(1-\alpha)\% = 99\%$$

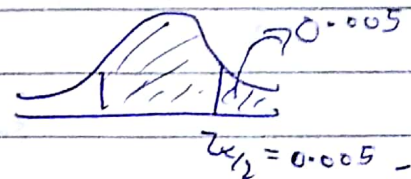
$$1-\alpha = 0.99$$

$$\alpha = 0.01 \Rightarrow \alpha/2 = 0.005$$

$$P(Z > z_{\alpha/2}) = 0.005$$

$$1 - P(Z > z_{\alpha/2}) = 1 - 0.005$$

$$P(Z < z_{0.005}) = 0.995 \Rightarrow z_{0.005} = 2.57$$



$$\bar{x} - \frac{s}{\sqrt{n}} z_{\alpha/2} < \mu < \bar{x} + \frac{s}{\sqrt{n}} z_{\alpha/2}$$

$$23500 - \frac{3900}{\sqrt{100}} \times 2.57 < \mu < 23500 + \frac{3900}{\sqrt{100}} \times 2.57$$

$$22,497.7 < \mu < 24,502.3$$

(6)

$$e < z_{\alpha/2} \frac{s}{\sqrt{n}}$$

$$e < 2.57 \times \frac{3900}{\sqrt{100}} = 1002.3$$

(3)

6.

✓

Q. 7.11: If \bar{x} is used as an estimate of μ , we can be $100(1-\alpha)\%$ confident that error will not exceed a specified amount e when sample size

$$n = \left(\frac{z_{\alpha/2} \sigma}{e} \right)^2$$

$$\sigma = 40, \quad z_{\alpha/2} = z_{0.02} = 2.05$$

$$e = 10$$

$$\therefore n = \left(\frac{z_{\alpha/2} \sigma}{e} \right)^2 = \left(\frac{2.05 \times 40}{10} \right)^2$$

$$n = 67.24$$

$$\boxed{n = 67}$$

7.

95% Confident.

$$\sigma = 0.0015, \quad e = 0.0005$$

$$1 - \alpha = 100(1-\alpha)\% = 95\%$$

$$\alpha = 0.05$$

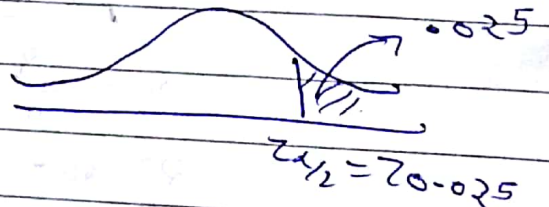
$$\Rightarrow \boxed{\alpha/2 = 0.025}$$

$$P(Z > z_{\alpha/2}) = \alpha/2$$

$$P(Z > z_{0.025}) = 0.025$$

$$P(Z < z_{0.025}) = 0.975$$

$$\therefore \boxed{z_{0.025} = 1.96}$$



$$n = \left(\frac{z_{\alpha/2} \sigma}{e} \right)^2 = \left(\frac{1.96 \times 0.0015}{0.0005} \right)^2$$

$$= \boxed{n = 34.57}$$

$$\boxed{n = 35}$$

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(230)

$$n = 10, \bar{x} = 230, s = 15$$

$$100(1-\alpha)\% = 99\%$$

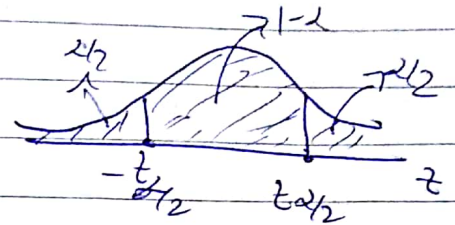
$$\alpha = 0.01 \quad \alpha/2 = 0.005$$

$$\text{deg. of freedom} = n-1 = 9$$

$$P(T > t_{\alpha/2}) = \alpha$$

$$t_{\alpha/2, n-1} = t_{0.005, 9} = 3.25$$

confidence interval for μ



$$\bar{x} - t_{\alpha/2, n-1} \frac{s}{\sqrt{n}} \leq \mu \leq \bar{x} + t_{\alpha/2, n-1} \frac{s}{\sqrt{n}}$$

$$230 - 3.25 \times \frac{15}{\sqrt{10}} \leq \mu \leq 230 + 3.25 \times \frac{15}{\sqrt{10}}$$

$$214.58 \leq \mu \leq 245.41$$

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Definition of Likelihood func:-

$$L(x_1, x_2, \dots, x_n, \theta) = f(x_1, \theta) \dots f(x_n, \theta)$$

$$L(x_1, \dots, x_n, \theta) = \prod_{i=1}^n f(x_i, \theta)$$

So Likelihood func:-

$$\theta = (\mu, \sigma^2)$$

$$L(x_1, \dots, x_n, \mu, \sigma^2) = \prod_{i=1}^n f(x_i, \mu, \sigma)$$

$$\text{Where } f(x, \mu, \sigma) = \begin{cases} \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{1}{2}\left(\frac{\ln x - \mu}{\sigma}\right)^2} & x > 0 \\ 0 & x < 0 \end{cases}$$

$$L(x_1, \dots, x_n, \mu, \sigma^2) = \prod_{i=1}^n \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{1}{2}\left(\frac{\ln x_i - \mu}{\sigma}\right)^2}$$

$$= \left(\frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{1}{2}\left(\frac{\ln x_1 - \mu}{\sigma}\right)^2} \right) \dots \left(\frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{1}{2}\left(\frac{\ln x_n - \mu}{\sigma}\right)^2} \right)$$

$$= \frac{1}{(\sqrt{2\pi})^n} \frac{1}{\sigma^n} \prod_{i=1}^n x_i e^{-\frac{1}{2}\left[\frac{(\ln x_1 - \mu)^2}{\sigma^2} + \dots + \frac{(\ln x_n - \mu)^2}{\sigma^2}\right]}$$

$$\ln\left(\frac{m}{n}\right) = \ln m - \ln n$$

$$= \frac{1}{(\sqrt{2\pi})^n} \frac{1}{\sigma^n} \prod_{i=1}^n x_i e^{-\frac{1}{2\sigma^2} \left[\sum_{i=1}^n (\ln x_i - \mu)^2 \right]}$$

$$\ln L = -\frac{1}{2\sigma^2} \left(\sum_{i=1}^n (\ln x_i - \mu)^2 \right) - n \ln(\sqrt{2\pi}) - n \ln \sigma^2 - \ln \prod_{i=1}^n x_i$$

$$\frac{\partial \ln L}{\partial \mu} = \frac{1}{\sigma^2} \cdot \sum_{i=1}^n (\ln x_i - \mu) \cdot (-1) = 0$$

$$\sum_{i=1}^n (\ln x_i - \mu) = 0$$

$$(\ln x_1 - \mu) + \dots + (\ln x_n - \mu) = 0$$

$$\sum_{i=1}^n \ln x_i - n\mu = 0$$

$$\mu = \frac{\sum_{i=1}^n \ln x_i}{n}$$

$$\text{So } \hat{\mu} = \frac{\sum_{i=1}^n \ln x_i}{n}$$

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$$\frac{\partial \ln L}{\partial \sigma^2} = \frac{-1}{2} \frac{(-1)}{(\sigma^2)^2} \left(\sum_{i=1}^n (\ln x_i - \mu)^2 \right) - \frac{n}{2} \frac{1}{\sigma^2} = 0$$

$$- \frac{n}{2 \sigma^2} + \frac{1}{2 \sigma^4} \sum_{i=1}^n (\ln x_i - \mu)^2 = 0$$

$$\frac{1}{2 \sigma^2} \sum_{i=1}^n (\ln x_i - \mu)^2 = \frac{n}{2}$$

$$\sigma^2 = \frac{\sum_{i=1}^n (\ln x_i - \mu)^2}{n}$$

$$\hat{\sigma}^2 = \frac{\sum_{i=1}^n (\ln x_i - \mu)^2}{n}$$

85.

$$L(x_1, \dots, x_n, \theta) = f(x_1, \theta) \cdot \dots \cdot f(x_n, \theta)$$

$$f(x, \theta) = f(x_0, \theta) = \begin{cases} \frac{1}{\theta} & 0 < x < \theta \\ 0 & \text{elsewhere} \end{cases}$$

$$L(x_1, \dots, x_n, \theta) = \underbrace{\frac{1}{\theta} \cdot \dots \cdot \frac{1}{\theta}}_{n \text{ times}} = \frac{1}{\theta^n}$$

$$\begin{cases} \text{here} \\ 0 < x_1 < \theta \\ 0 < x_2 < \theta \\ \vdots \\ 0 < x_n < \theta \end{cases}$$

∴ Likelihood function of θ

$$L(x_1, \dots, x_n, \theta) = \frac{1}{\theta^n}, \text{ for } \max\{x_1, \dots, x_n\} < \theta < \infty$$

$$\text{ie. } \begin{matrix} x_1 < \theta \\ x_2 < \theta \\ \vdots \\ x_n < \theta \\ \max\{x_1, \dots, x_n\} < \theta \end{matrix}$$