

From last class,

For LRT's of 1-parameter: $H_0: \mu = \mu_0$

To find $100(1-\alpha)\%$ C.I for μ ,

find set of μ_0 , such that,

$$p\text{-value} = P[D \geq d_{\text{obs}}(\mu_0) \mid H_0 \text{ true}] \geq \alpha$$

Under H_0 , $D \approx \chi^2_{(1)}$. Hence,

$$P[D \geq d_{\text{obs}}(\mu_0) \mid H_0 \text{ true}] \geq \alpha \Leftrightarrow d_{\text{obs}}(\mu_0) \leq \chi^2_{\alpha, (1)}$$

Critical values comes from your test
Statistic's sampling distribution..

Critical Value.

Section 5.2 \Rightarrow Approximating C.I's (with Likelihood intervals)

$$H_0: \mu = \mu_0$$

Recall under H_0 , LRS $D \equiv -2\ln(\mu) \approx \chi^2_{(1)}$

For a significance level α ,

$$d_{\text{obs}}(\mu_0) \leq \chi^2_{\alpha, (1)}$$

characteristics of
100(1- α)% C.I.

$$\Rightarrow -2\ln(\mu_0) \leq \chi^2_{\alpha, (1)}$$

$$\Rightarrow \ln(\mu_0) \geq -\frac{1}{2} \chi^2_{\alpha, (1)}$$

$$\Rightarrow R(\mu_0) \geq e^{-\frac{1}{2} \chi^2_{\alpha, (1)}}$$

characteristics of
100 \times $e^{-\frac{1}{2} \chi^2_{\alpha, (1)}}$ % L.I

<u>C.I %</u>	<u>Equivalent L.I %</u>
90	25.8
95	14.7
96	10
99	3.6

Back to our previous Example

$$P[X_{(1)}^2 \geq 3.843] = 0.05$$

$$\uparrow D = -2\eta(\mu) = 3.843$$

$$\Rightarrow \eta(\mu) = -1.9215$$

$$\Rightarrow e^{\eta(\mu)} = e^{-1.9215}$$

$$\Rightarrow R(\mu) = 0.147 \quad \left. \begin{array}{l} 14.7\% \text{ L.S.} \\ = 95\% \text{ C.I.} \end{array} \right\}$$

$$X_{(1)}^2 = 3.843$$

From Table

then,

$$R(\mu) = e^{-\frac{1}{2}(3.843)}$$

★ Convenient when you have R ~~available~~ available to compute L.S.'s

$$R(\mu) \leftrightarrow [A, B] \leftrightarrow \alpha$$

Section 5.3 \rightarrow Another Approximation for C.I.'s.

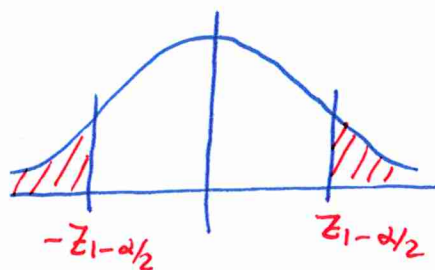
In many cases (all of them, in this course), MLE's have an approx. Normal distribution for large sample sizes n .

$$\hat{\theta} \approx N(\theta_0, \frac{1}{\sqrt{I(\hat{\theta})}}), \quad \theta_0 \text{ is the true value.}$$

$$\Leftrightarrow \frac{\hat{\theta} - \theta_0}{\sqrt{I(\hat{\theta})^{-1}}} \text{ or } \frac{\hat{\theta} - \theta_0}{1/\sqrt{I(\hat{\theta})}} \text{ or } \sqrt{I(\hat{\theta})}(\hat{\theta} - \theta_0) \approx N(0, 1)$$

$$100(1 - \alpha)\% \text{ C.I. : } \hat{\theta} \pm \frac{Z_{1-\alpha/2}}{\sqrt{I(\hat{\theta})}},$$

$$H_0: \theta = \theta_0, \quad H_1: \theta \neq \theta_0 \quad \leftarrow \text{2 tailed test.}$$



$$Z_{\alpha/2} = -Z_{1-\alpha/2}$$

Now, Let's look at an Example,
Example 5.2.1 from the Complete Lecture Notes.

Here, $X_{\text{obs}} = 23$.

$$X \sim \text{Binomial}(n=100, \theta)$$

$$\text{MLE of } \theta, \hat{\theta} = 23/100 = 0.23$$

Method-1: $H_0: \theta = \theta_0$
 $X \sim \text{Bin}(100, \theta_0)$, under H_0

If, $n\theta_0 > 5$ and $n(1-\theta_0) > 5$,

Then $X \approx \text{Normal}(\mu = n\theta_0, \sigma^2 = n\theta_0(1-\theta_0))$

$$\text{Let, } D = \left| \frac{X - n\theta_0}{\sqrt{n\theta_0(1-\theta_0)}} \right| = |Z|, \quad Z \sim N(0,1)$$

$$P(D \geq d_{\text{obs}}) = P(|Z| \geq |Z_{\text{obs}}|) \geq 0.05$$

$$\Rightarrow P(|Z| \geq \left| \frac{23 - 100\theta_0}{10\sqrt{\theta_0(1-\theta_0)}} \right|) \geq 0.05$$

$$\Leftrightarrow \left| \frac{23 - 100\theta_0}{10\sqrt{\theta_0(1-\theta_0)}} \right| \leq Z_{0.975} = 1.96$$

Critical values
from
table

① Solve for θ_0 .

② Approximate $\sqrt{\theta_0(1-\theta_0)}$ with $\sqrt{\hat{\theta}(1-\hat{\theta})}$.

$$\text{Using (2), } \frac{23}{100} \pm 1.96 \sqrt{\frac{0.23(1-0.23)}{10}}$$

$$= [0.148, 0.312]$$

Method-2: 95% C.I. = 14.7% L.I. = $[0.155, 0.319]$

Method-3 : $\hat{\theta} \approx \mathcal{N}(\theta_0, \mathcal{I}(\hat{\theta})^{-1})$

$$L(\theta) = \theta^x (1-\theta)^{n-x}$$

$$\ell(\theta) = x \ln(\theta) + (n-x) \ln(1-\theta)$$

$$\ell'(\theta) = \frac{x}{\theta} - \frac{n-x}{1-\theta}$$

$$\ell''(\theta) = -\frac{x}{\theta^2} - \frac{n-x}{(1-\theta)^2}$$

$$\mathcal{I}(\theta) = -\ell''(\theta) = \frac{x}{\theta^2} + \frac{(n-x)}{(1-\theta)^2}$$

$$\mathcal{I}(\hat{\theta}) = \frac{\cancel{x} \cdot n^2}{\cancel{x}^2} + \frac{n-x}{(1-\cancel{x}/n)^2} = \frac{n^2}{x} + \frac{(n-x)n^2}{(n-x)^2}$$

$$\Rightarrow \mathcal{I}(\hat{\theta}) = \frac{n}{\hat{\theta}} + \frac{n}{(1-\hat{\theta})} \quad \dots \quad [\hat{\theta} = x/n]$$

$$\mathcal{I}(\hat{\theta})^{-1} = \left(\frac{n}{\hat{\theta}} + \frac{n}{(1-\hat{\theta})} \right)^{-1} \quad \left[\begin{array}{l} \text{* Need common denominators} \\ \text{before you can take reciprocal!} \end{array} \right]$$

$$= \left(\frac{n(1-\hat{\theta}) + n\hat{\theta}}{\hat{\theta}(1-\hat{\theta})} \right)^{-1} = \left(\frac{n}{\hat{\theta}(1-\hat{\theta})} \right)^{-1}$$

$$\Rightarrow \mathcal{I}(\hat{\theta})^{-1} = \hat{\theta}(1-\hat{\theta})/n$$

Therefore, $\hat{\theta} \pm 1.96 \sqrt{\frac{\hat{\theta}(1-\hat{\theta})}{n}}$

$$= [0.148, 0.312] \quad \text{in the 95\% C.I.}$$
