

Sketch proof for univariate θ

Given that we are discussing ML under suitable regularity conditions,

$$\frac{dL}{d\theta}(\hat{\theta}) = 0 \quad (L'(\hat{\theta}) = 0)$$

(Now further assume X_1, \dots, X_n iid say later when needed)
Taylor-expand L' about the "true" value θ , so that

$$0 = L'(\hat{\theta}) \approx L'(\theta) + (\hat{\theta} - \theta) L''(\theta)$$

$$\Rightarrow (\hat{\theta} - \theta) \approx \frac{-L'(\theta)}{L''(\theta)}$$

$$\sqrt{n}(\hat{\theta} - \theta) \approx \frac{-\frac{1}{\sqrt{n}} L'(\theta)}{\frac{1}{n} L''(\theta)}$$

Consider first the RHS numerator

$$\frac{1}{\sqrt{n}} L'(\theta) = \frac{1}{\sqrt{n}} \sum_{i=1}^n \frac{d \ln f_{\theta}(x_i)}{d\theta} \quad \text{as indep}$$

We showed earlier that $E\left(\frac{d \ln f_{\theta}(x)}{d\theta}\right) = 0$ and $\text{Var}\left(\frac{d \ln f_{\theta}(x)}{d\theta}\right) = I_1(\theta)$ (8.2.4), so we

have the sum of n iid random variables, which by the CLT converges to a Normal

$$-\frac{1}{\sqrt{n}} \sum_{i=1}^n \frac{d \ln f_{\theta}(x_i)}{d\theta} \xrightarrow{\substack{S_n - n\mu \\ \sqrt{n\sigma^2}}} N(0, I_1(\theta)) \xrightarrow{D} N(0, 1)$$

Now let's consider the RHS denominator

$$\frac{1}{n} \sum_{i=1}^n \frac{d^2 \ln f_{\theta}(x_i)}{d\theta^2} \quad \frac{1}{n} \sum_{i=1}^n X_i \rightarrow \mu$$

Again we have the sum of n random variables

which are iid with mean $-\dot{I}_1(\theta)$. Since we are considering their sample mean $\frac{1}{n} \sum_1^n \frac{d^2 \ln f_\theta(x_i)}{d\theta^2}$ the strong law of large numbers tells us that

$$-\frac{1}{n} \sum_1^n \frac{d^2 \ln f_\theta(x_i)}{d\theta^2} \rightarrow \dot{I}_1(\theta)$$

Putting these two results together (Slutsky's thm)

$$\sqrt{n} (\hat{\theta} - \theta) \rightarrow \frac{1}{\dot{I}_1(\theta)} N(0, \dot{I}_1(\theta))$$

$$\text{or } (\hat{\theta} - \theta) \rightarrow N\left(0, \frac{1}{n \dot{I}_1(\theta)}\right)$$

$$\text{ie } \hat{\theta} \sim N\left(\theta, \frac{1}{n \dot{I}_1(\theta)}\right) \text{ asymptotically}$$