

Information From a View of MLE Asymptotic Distribution

Denote $L(x, \theta)$ as the likelihood function.

And let $l(x, \theta) = \ln(L(x, \theta))$ as the log-likelihood.

$$1. E\left[\frac{\partial l}{\partial \theta}\right] = 0 \quad \frac{\partial l}{\partial \theta} = \frac{1}{L} \frac{\partial L}{\partial \theta} \quad \text{constant 1.}$$

$$\int_x \frac{1}{L} \frac{\partial L}{\partial \theta} \cdot L \, dx = \int_x \frac{\partial L}{\partial \theta} \, dx = \frac{\partial}{\partial \theta} \int_x L \, dx = 0$$

$$2. \text{Var}\left[\frac{\partial l}{\partial \theta}\right] = -E\left[\frac{\partial^2 l}{\partial \theta^2}\right] = R(\theta)$$

Fisher Information

$$0 = \frac{\partial}{\partial \theta} E\left(\frac{\partial l}{\partial \theta}\right) = \frac{\partial}{\partial \theta} \int_x \frac{\partial l}{\partial \theta} L \, dx = \int_x \frac{\partial^2 l}{\partial \theta^2} L \, dx + \int_x \frac{\partial l}{\partial \theta} \frac{\partial L}{\partial \theta} \cdot \frac{1}{L} \cdot L \, dx$$

$$= -R(\theta) + \text{Var}\left[\frac{\partial l}{\partial \theta}\right]$$

$$3. \frac{\partial l}{\partial \theta}(x; \theta) \stackrel{\text{Taylor expansion}}{=} \frac{\partial l}{\partial \theta}(x; \theta_0) + (\theta - \theta_0) \frac{\partial^2 l}{\partial \theta^2}(x; \theta_0) + \frac{1}{2} (\theta - \theta_0)^2 \frac{\partial^3 l}{\partial \theta^3}(x; \theta_0)$$

$$\uparrow \quad \quad \quad \uparrow$$

$$n\beta_0 \quad \quad \quad n\beta_1$$

$$\beta_0 \xrightarrow{P} 0 \quad \text{From ①} \quad \beta_1 \xrightarrow{P} R(\theta) \quad \text{From ②}$$

$$n\beta_0 + (\theta - \theta_0)n\beta_1 + \frac{1}{2}(\theta - \theta_0)^2 \frac{\partial^3 l}{\partial \theta^3}(\theta_0) = 0$$

$$\beta_0 + (\theta - \theta_0)\beta_1 + \frac{1}{2}(\theta - \theta_0)^2 \frac{\partial^3 l}{\partial \theta^3}(\theta_0) = 0$$

$$(\theta - \theta_0) \left(\beta_1 + \frac{1}{2}(\theta - \theta_0) \frac{\partial^3 l}{\partial \theta^3}(\theta_0) \right) = -\beta_0$$

$$\sqrt{n}(\hat{\theta} - \theta_0) \left(\beta_1 + \frac{1}{2}(\hat{\theta} - \theta_0) \frac{\partial^3 l}{\partial \theta^3}(\theta_0) \right) = \frac{1}{\sqrt{n}} \sum \frac{\partial^2 l}{\partial \theta^2}(x_i; \theta_0)$$

$$\downarrow R(\theta) \quad \quad \quad \downarrow N(0, R(\theta))$$

$$\sqrt{n R(\theta)} (\hat{\theta} - \theta_0) \xrightarrow{P} N(0, 1)$$

$$\sqrt{n} (\hat{\theta} - \theta_0) \xrightarrow{P} N(0, \frac{1}{R(\theta)})$$

$$4. -\frac{1}{\sqrt{n}} l'(\theta) \xrightarrow{d} N(\theta, I(\theta))$$

$$\frac{1}{n} l''(\theta) \xrightarrow{P} I(\theta)$$