

2003 Fall Part 1

1. Fisher Information

$$\log f_X(x) = -\theta + x \log \theta - \log x!$$

$$\begin{aligned}\frac{\partial \log f_X(x)}{\partial \theta} &= -1 + \frac{x}{\theta} \\ \frac{\partial^2 \log f_X(x)}{\partial \theta^2} &= -\frac{x}{\theta^2}\end{aligned}$$

Therefore, Fisher Information is

$$I(\theta) = -E \left[\frac{\partial^2 \log f_X(x)}{\partial \theta^2} \right] = \frac{E[X]}{\theta^2} = \frac{1}{\theta}$$

2. Asymptotic distribution

Let $g(x) = x^2$, then $g(1) = 1$ and $g'(1) = 2$. Apply delta method,

$$\begin{aligned}\sqrt{n}(g(Z_n) - g(1)) &\xrightarrow{d} N(0, [g'(1)]^2 \cdot 4) \\ \sqrt{n}(Z_n^2 - 1) &\xrightarrow{d} N(0, 16)\end{aligned}$$

3. Summation of independent normal distribution

MGF of $\sum_i a_i X_i$ is

$$\begin{aligned}E \left[e^{t \sum_i a_i X_i} \right] &= E \left[e^{ta_1 X_1 + ta_2 X_2 + \dots + ta_n X_n} \right] \\ &= E \left[e^{ta_1 X_1} e^{ta_2 X_2} \dots e^{ta_n X_n} \right] \\ &= E \left[e^{ta_1 X_1} \right] E \left[e^{ta_2 X_2} \right] \dots E \left[e^{ta_n X_n} \right] \quad \because \text{independence} \\ &= \exp \left(\mu_1 ta_1 + \frac{1}{2} \sigma_1^2 (ta_1)^2 \right) \exp \left(\mu_2 ta_2 + \frac{1}{2} \sigma_2^2 (ta_2)^2 \right) \dots \exp \left(\mu_n ta_n + \frac{1}{2} \sigma_n^2 (ta_n)^2 \right) \\ &= \exp \left([a_1 \mu_1 + a_2 \mu_2 + \dots + a_n \mu_n] t + \frac{1}{2} [a_1^2 \sigma_1^2 + a_2^2 \sigma_2^2 + \dots + a_n^2 \sigma_n^2] t^2 \right)\end{aligned}$$

This is MGF of normal distribution with mean $\sum_i a_i \mu_i$ and variance $\sum_i a_i^2 \sigma_i^2$.

4. Method of Moments (a)

$$E[X_i^2] = \text{var}(X_i) + (E[X_i])^2 = \theta + \theta^2$$

(b)

Substitute moment condition with the sample moment condition

$$\begin{aligned}\frac{1}{100} \sum_{i=1}^{100} X_i^2 &= \theta + \theta^2 \\ 6 &= \theta + \theta^2\end{aligned}$$

Solving the equation, $\theta = 2$ or -3 . Since $\theta = E[X_i] = \text{var}(X_i) > 0$,

$$\hat{\theta}_{MM} = 2$$