

PH 712 Probability and Statistical Inference

Part VIII: Point Estimation II (Aymptotic Properties)

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Consistency of MM and ML estimators

- For an iid sample, under certain conditions:
 - $\hat{\theta}_{\text{MM}} \approx \theta$ as $n \rightarrow \infty$
 - $\hat{\theta}_{\text{ML}} \approx \theta$ as $n \rightarrow \infty$

Asymptotic efficiency of MLE (CB Thm 10.1.12 & Ex. 10.7)

- For an iid sample, under certain conditions:
 - $\sqrt{n}(\hat{\theta}_{\text{ML}} - \theta) \approx \mathcal{N}(0, I_1^{-1}(\theta))$ as $n \rightarrow \infty$
 - * For an iid sample, $I_1(\theta) = n^{-1}I_n(\theta)$, no longer a function of n
 - In practice, unknown $\theta \Rightarrow$ unknown $I_n(\theta)$
 - * $I_n(\theta) \approx I_n(\hat{\theta}_{\text{ML}}) \approx \hat{I}_n(\hat{\theta}_{\text{ML}})$
 - Fisher information (evaluated at θ) $I_n(\theta) = -E\{H(\theta)\}$
 - Observed Fisher information $\hat{I}_n(\hat{\theta}_{\text{ML}}) = -H(\hat{\theta}_{\text{ML}})$
 - Application (approximating the distribution of $\hat{\theta}_{\text{ML}}$): $\hat{\theta}_{\text{ML}} \approx \mathcal{N}(\theta, \hat{I}_n^{-1}(\hat{\theta}_{\text{ML}}))$, i.e., approximately, $\hat{\theta}_{\text{ML}}$ is normally distributed with mean θ and variance $\hat{I}_n^{-1}(\hat{\theta}_{\text{ML}})$.

Delta method

- (CB Thm 5.5.24, delta method) Given an estimator T_n , if $\sqrt{n}(T_n - \theta) \approx \mathcal{N}(0, \sigma^2)$, h is NOT a function of n , AND $h'(\theta) \neq 0$, then

$$\sqrt{n}\{h(T_n) - h(\theta)\} \approx \mathcal{N}(0, \{h'(\theta)\}^2 \sigma^2).$$

– Hence $\text{var}\{h(T_n)\} \approx \{h'(\theta)\}^2 \sigma^2 / n$ if $h'(\theta) \neq 0$.

- (CB Thm 5.5.26, second-order delta method) Given an estimator T_n , if $\sqrt{n}(T_n - \theta) \approx \mathcal{N}(0, \sigma^2)$, h is NOT a function of n , $h'(\theta) = 0$, AND $h''(\theta) \neq 0$, then

$$n\{h(T_n) - h(\theta)\} \approx \frac{h''(\theta)\sigma^2}{2}\chi^2(1).$$

– Hence $\text{var}\{h(T_n)\} \approx \{h''(\theta)\}^2 \sigma^4 / (2n^2)$ if $h'(\theta) = 0$ but $h''(\theta) \neq 0$.

CB Example 10.1.17 & Ex. 10.9

- $X_1, \dots, X_n \stackrel{\text{iid}}{\sim} p(x | \lambda) = \lambda^x \exp(-\lambda) / x!$, $x \in \mathbb{Z}^+$, $\lambda > 0$. To estimate $\Pr(X_i = 0) = \exp(-\lambda)$.
 1. What is the MLE for $\Pr(X_i = 0)$, say W_n ?
 2. Approximate the variance of W_n .
 3. Suppose $T_n = n^{-1} \sum_i \mathbf{1}_{\{0\}}(X_i)$. Approximate the variance of T_n .
 4. Compute $\text{ARE}(T_n, W_n)$, the ARE of T_n with respect to W_n .