

ASYMPTOTIC VARIANCE of the MLE

Maximum likelihood estimators typically have good properties when the sample size is large.

Suppose X_1, \dots, X_n are iid from some distribution F_{θ_o} with density f_{θ_o} . We observe data x_1, \dots, x_n .

The Likelihood is:

$$L(\theta) = \prod_{i=1}^n f_{\theta}(x_i)$$

and the log likelihood is:

$$l(\theta) = \sum_{i=1}^n \log[f_{\theta}(x_i)]$$

The maximum likelihood estimator maximizes $l(\cdot)$, and is an estimator for θ_o . Under some regularity conditions of f_{θ} , for n large

$$\hat{\theta} \approx N\left(\theta_o, \frac{1}{nI(\theta_o)}\right)$$

where $I(\theta)$ is called the *information*, and is defined as

$$I(\theta) = E\left(\frac{\partial \log f_{\theta}(X)}{\partial \theta}\right)^2$$

Notice that X is capitalized above. It denotes that the expectation is being taken with respect to X and its distribution. Also note that the derivative is with respect to θ .

It is often the case that the information also can be expressed as

$$I(\theta) = -E\left(\frac{\partial^2 \log f_{\theta}(X)}{\partial \theta^2}\right)$$

For $\tilde{\theta}$ any unbiased estimator for θ_o , we have a lower bound on the variance of $\tilde{\theta}$

$$\text{Var}(\tilde{\theta}) \geq \frac{1}{nI(\theta_o)}$$

Let's work some examples.

Example: X_1, \dots, X_n iid with density $f_\sigma(x) = \frac{1}{2\sigma} \exp(-|x|/\sigma)$.

Method of Moments

1. $E(X) = 0$ by symmetry

$$E(X^2) = \int_{-\infty}^{\infty} x^2 f_\sigma(x) dx =$$

2. $\sigma =$

3. $\hat{\sigma} =$

Method of Moments

1. Likelihood function: $L(\sigma) =$

$$\log \text{likelihood } l(\sigma) =$$

2. Differentiate with respect to σ

$$\frac{\partial \log l(\sigma)}{\partial \sigma} =$$

3. Set to 0 and solve for $\tilde{\sigma}$

Asymptotic Variance of $\tilde{\sigma}$

$$\frac{\partial \log f_{\sigma}(x)}{\partial \sigma} =$$

$$\frac{\partial^2 \log f_{\sigma}(x)}{\partial \sigma^2} =$$

$$I(\sigma) = -E\left[\frac{\partial^2 \log f_{\sigma}(x)}{\partial \sigma^2}\right] =$$

The variance of $\tilde{\theta}$ is asymptotically equivalent to

$$\frac{1}{nI(\sigma)} =$$

How would we make a confidence interval for σ using $\hat{\sigma}$ and this asymptotic variance?

Example: X_1, \dots, X_n iid with density $f_\theta(x) = (\theta + 1)x^\theta$ for $0 \leq \theta \leq 1$.

Method of Moments

1. $E(X) = \int_0^1 (\theta + 1)x^{\theta+1} dx$

2. $\theta =$

3. $\hat{\theta} =$

Method of Moments

1. Likelihood function: $L(\theta) =$

log likelihood $l(\theta) =$

2. Differentiate with respect to θ

$$\frac{\partial \log l(\theta)}{\partial \theta} =$$

3. Set to 0 and solve for $\tilde{\theta}$

Asymptotic Variance of $\tilde{\theta}$

$$\frac{\partial \log f_{\theta}(x)}{\partial \theta} =$$

$$\frac{\partial^2 \log f_{\theta}(x)}{\partial \theta^2} =$$

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

The variance of $\tilde{\theta}$ is asymptotically equivalent to

$$\frac{1}{nI(\theta)} =$$