

STAT 703/J703
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-Lecture 7-

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Today

- Consistency of the MLE
- Information Function
- Asymptotic Normality of the MLE

8.5.2 – Large Sample Properties

Theorem A: The MLE is Consistent
(under appropriate regularity conditions)

Sketch of Proof: Consider

$$\text{maximizing } \frac{1}{n} L(\theta) = \frac{1}{n} \sum_{i=1}^n \log f(X_i | \theta)$$

Information function

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



Example: Consider a random sample of size n from a normal distribution with unknown mean θ and known variance σ^2 .



What if there is more than one parameter?

In this case you get an information matrix:

$$I(\theta) = E \left[\left(\frac{\partial}{\partial \theta_i} \log f(X | \theta) \right) \left(\frac{\partial}{\partial \theta_j} \log f(X | \theta) \right) \right]$$
$$= -E \left[\frac{\partial^2}{\partial \theta_i \partial \theta_j} \log f(X | \theta) \right]$$



Theorem B: Under appropriate regularity conditions the MLE is asymptotically normal with mean θ and variance $\frac{1}{nI(\theta)}$.

Sketch of proof: Consider the Taylor series expansion:

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

Example: Recall the multinomial example from section 8.5.1.
