

Lecture 13: Method of Moments

- To estimate p parameters solve

$$\bar{X} = E_{\theta}(X)$$

$$\bar{X}^2 = E_{\theta}(X^2)$$

$$\vdots \quad \vdots$$

$$\bar{X}^p = E_{\theta}(X^p)$$

- Essentially always consistent.
- Did Gamma(α, β) example.
- Used as starting points for Newton-Raphson.
- To solve $g(\theta) = 0$ begin with initial value θ_0 and iteratively define

$$\theta^{(k+1)} = \theta^{(k)} - \left(Dg(\theta^{(k)}) \right)^{-1} g(\theta^{(k)}).$$

Here Dg is the $p \times p$ matrix with i, j th

$$\frac{\partial g_i(\theta)}{\partial \theta_j}.$$



Lecture 13: quadratic approximation to ℓ

- Two term Taylor expansion of ℓ about $\hat{\theta}$:

$$2 \left[\ell(\hat{\theta}) - \ell(\theta) \right] \approx (\theta - \hat{\theta})^T \mathcal{I}(\theta_0) (\theta - \hat{\theta})$$

where minus the second derivative matrix has been replaced by the Fisher information matrix.

- Worked on large sample theory of above evaluated at $\theta = \theta_0$.
- Began with $X \sim MVN_p(0, \Sigma)$ and showed

$$X^T \Sigma^{-1} X \sim \chi_p^2.$$

- Then started general theory of quadratic form:

$$X^t B X = (AZ)^T B (AZ) = Z^T Q Z$$

where $Q = A^T B A$ and $A A^T = \Sigma$.

- Write $Q = P \Lambda P^T$ where P is orthogonal ($P P^T = P^T P = I$) and Λ is diagonal. Columns of P are eigenvectors of Q and diagonal entries Λ are eigenvalues.



Coverage in the text

- Method of moments in Chapter 9.
- See “course notes” on web pages 130-131.

