

## Lecture 8: Likelihood Asymptotics

- Suppose  $X_1, \dots, X_n$  iid with density  $f_\theta(x)$ . Parameter  $\theta \in \Theta \subset \mathbb{R}^p$ .
- *Likelihood function* is

$$L(\theta) = \prod_1^n f_\theta(X_i).$$

- The *log-likelihood function* is

$$\ell(\theta) = \sum_1^n \log f_\theta(X_i).$$

- The *Score function* is the gradient of  $\ell$ :

$$U(\theta) = \nabla \ell(\theta) = \frac{\partial \ell}{\partial \theta} = \sum_1^n \frac{\partial \log f_\theta(X_i)}{\partial \theta}$$

- The Hessian is the  $p \times p$  matrix with  $jk$ th entry

$$V_{jk}(\theta) = \sum_{i=1}^n \frac{\partial^2 \log f_\theta(X_i)}{\partial \theta_j \partial \theta_k}$$

- The *Fisher Information* matrix is



# Bartlett identities

- First Bartlett identity

$$E_\theta(U(\theta)) = 0$$

- Second Bartlett identity

$$\mathcal{I}(\theta) = -E_\theta(V(\theta)) = \text{Var}_\theta(U(\theta)).$$

- Look up properties of variance and covariance.



# Course coverage

- Chapter 9.3-7.

