

Repetition week 44

The mean square error

$$MSE = E[(W - \theta)^2] = Var[W] + (E[W] - \theta)^2$$

Score function

$$S(\mathbf{X}|\theta) = \frac{\partial}{\partial \theta} \log f(\mathbf{X}|\theta)$$

$$E[S(\mathbf{X}|\theta)] = 0$$

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

$$\text{Let } \tau(\theta) = E[W(\mathbf{X})]$$

Cramer-Rao

$$Var[W(\mathbf{X})] \geq \frac{\left(\frac{\partial}{\partial \theta} \tau(\theta)\right)^2}{I_{\mathbf{X}}(\theta)}$$

Cramer-Rao iid

$$\text{Var}[W(\mathbf{X})] \geq \frac{\left(\frac{\partial}{\partial \theta} \tau(\theta)\right)^2}{nI_X(\theta)}$$

Equality

If and only if $S(\mathbf{X}|\theta) = a(\theta)[W(\mathbf{X}) - b(\theta)]$

The multiparameter case

$$\text{Var}[W(\mathbf{X})] \geq (\nabla \tau(\theta))^T (I(\theta))^{-1} (\nabla \tau(\theta)).$$

Definition 6.2.21

Let $f(t|\theta)$ be a family of pdfs/pmfs for a statistic $T(\mathbf{X})$. The family is complete if

$$E_\theta[g(T)] = 0 \Rightarrow P_\theta(g(T) = 0) = 1, \text{ for all } \theta.$$

Theorem 6.2.25:

Let X_1, \dots, X_n be iid. from an exponential family i.e.

$$f(x|\boldsymbol{\theta}) = h(x)c(\boldsymbol{\theta})e^{\sum_{i=1}^k w(\theta_i)t_i(x)}$$

Then $T(\mathbf{X}) = \left(\sum_{i=1}^n t_1(X_i), \sum_{i=1}^n t_2(X_i), \dots, \sum_{i=1}^n t_k(X_i)\right)$ is complete as long

as the parameter space contains an open set in R^n .

