

Stat 610

Final Examination

Due Dec. 9, 2005, 2:30PM

1 (20 points). Let X_1, X_2, \dots, X_n be i.i.d. according to $Cauchy(\theta, 1)$. Assume that Pitman estimator δ_P is an asymptotically efficient estimator of θ with $\lim_{n \rightarrow \infty} var(\sqrt{n}\delta_P) = 2$. Find an unbiased estimator of θ^3 such that it is asymptotically efficient at $\theta \neq 0$.

Hint: you may assume the third moment of δ_P is finite.

2 (20 points). Let

$$X_i | \theta \stackrel{i.i.d.}{\sim} U(0, \theta), \quad i = 1, 2, \dots, n$$

and

$$\frac{1}{\theta} | a, b \sim Gamma(a, b), \quad a, b \text{ known}$$

Show that

- (i) The Bayes estimator will only depend on the data through $Y = \max_i X_i$.
- (ii)

$$E(\theta | a, b) = \frac{1}{b(n+a-1)} \frac{P\left(\chi_{2(n+a-1)}^2 < 2/(by)\right)}{P\left(\chi_{2(n+a)}^2 < 2/(by)\right)}$$

3 (20 points). Let $X \sim Bernoulli(p)$ with unknown $1 > p > 0$.

- (i) Is X a minimax estimator of p under the squared error loss $L(p, \delta) = (p - \delta)^2$. Why?
- (ii) Is X an admissible estimator of p under the squared error loss $L(p, \delta) = (p - \delta)^2$. Why?

4 (20 points). Observe $X = (X_1, X_2, \dots, X_n)$ be $N(\theta, I_n)$ with unknown $\theta = (\theta_1, \theta_2, \dots, \theta_n) \in R^n$. For given $p \geq 1$ and $\lambda > 0$, define

$$\delta_i(X) = \left(1 - \frac{\lambda^p}{|X_i|^p}\right)_+ X_i.$$

Show that

- (i)

$$E_{X|\theta} \sum_{i=1}^n (\delta_i(X) - \theta_i)^2 = E_{X|\theta} SURE(\lambda)$$

with

$$SURE(\lambda) = n + \sum_{i=1}^n \left[(X_i^2 - 2) I(|X_i| \leq \lambda) + \left(\frac{2(p-1)\lambda^p}{|X_i|^p} + \frac{\lambda^{2p}}{|X_i|^{2p-2}} \right) I(|X_i| > \lambda) \right]$$

(ii) The minimum of $SURE(\lambda)$ is achieved at some $\lambda \in A$ with

$$A = \{|X_i|; 1 \leq i \leq n\} \cup \{0\}.$$

5 (20 points). Let X_1, X_2, \dots, X_n be i.i.d. with density

$$f_{\theta, \sigma}(x) = \frac{\theta}{\sigma} x^{\theta-1} e^{-x^\theta/\sigma}, \quad x > 0, \theta > 0, \sigma > 0.$$

where θ and σ are unknown. Show that

- (i) The likelihood equation has a unique solution unless all X_i are equal.
- (ii) The sequence of maximum likelihood estimators $(\hat{\theta}_{MLE}, \hat{\sigma}_{MLE})$ is asymptotically normal.