# UC Berkeley Department of Statistics

#### STAT 210A: Introduction to Mathematical Statistics

## Problem Set 5

Fall 2014

Issued: Friday, October 10 Due: Thursday, October 16 (beginning of class)

### Problem 5.1

If  $X_1, \ldots, X_n$  are i.i.d. from  $N(\theta, \theta)$ , then two natural estimators of  $\theta$  are the sample mean  $\bar{X}$  and the sample variance  $S^2$ . Determine the asymptotic relative efficiency of  $S^2$  with respect to  $\bar{X}$ .

#### Problem 5.2

Let  $X_1, \ldots, X_n$  be i.i.d. from an exponential distribution with unit failure rate.

- 1. Suppose we are interested in the limiting distribution for  $X_{(2)}$ , the second order statistic. Naturally,  $X_{(2)} \stackrel{p}{\to} 0$  as  $n \to \infty$ . For an interesting limit theory we should scale  $X_{(2)}$  by an appropriate power of n, but the correct power is not 1/2. Suppose x > 0. Find a value p so that  $P(n^p X_{(2)} \le x)$  converges to a value between 0 and 1. (If p is too small, the probability will tend to 1, and if p is too large the probability will tend to 0.)
- 2. Determine the limiting distribution for  $X_{(n)} \log n$ .

#### Problem 5.3

Consider the loss function

$$L(\theta, a) = \begin{cases} k_1 & |\theta - a| & \text{if } a \le \theta \\ k_2 & |\theta - a| & \text{if } a > \theta \end{cases}$$

where  $k_1 > 0$  and  $k_2 > 0$  are constants. In a Bayesian setting, suppose that the random variable  $(\theta \mid X = x)$  has finite mean for each x. Show that under this loss function, Bayes estimators are  $p^{th}$  quantiles of the posterior distribution, where p is a suitable function of  $k_1$  and  $k_2$ .

#### Problem 5.4

Given a fixed known integer r > 1, let  $X_{ij}$ , j = 1, ..., r and i = 1, ..., n be i.i.d. samples from  $N(\mu_i, \sigma^2)$ . Find the MLE of  $\theta = (\mu_1, ..., \mu_n, \sigma^2)$ , and show that it is inconsistent for  $\sigma^2$  as  $n \to +\infty$ .

### Problem 5.5

Let  $(X_1, \ldots, X_n)$  be an i.i.d. sample from the mixture distribution with density

$$f_{\theta}(x) = \theta f_1(x) + (1 - \theta) f_2(x),$$

where  $f_i$ , i = 1, 2 are two different known densities, and  $\theta \in (0, 1)$  is unknown.

(a) Show that the conditions

$$\frac{1}{n} \sum_{i=1}^{n} \frac{f_1(X_i)}{f_2(X_i)} > 1$$
 and  $\frac{1}{n} \sum_{i=1}^{n} \frac{f_2(X_i)}{f_1(X_i)} > 1$ 

are necessary and sufficient for the score equation (setting the derivative of the log likelihood to zero) to have a unique solution. Show that if there is a solution, then it is the MLE.

(b) Derive the MLE of  $\theta$  when the score equation has no solution.