# UC Berkeley Department of Statistics Fall 2016

### STAT 210A: Introduction to Mathematical Statistics

### Problem Set 10

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Note: All measure-theoretic niceties about conditioning on measure-zero sets, almost-sure equality vs. actual equality, "all functions" vs. "all measurable functions," etc. are disregarded for the time being.

## 1. Uniform Convergence

Take K = [0, 1], let  $W_n$ ,  $n \ge 1$ , be random functions taking values in C(K), and let f be a fixed function in C(K). Consider the following conjecture: If  $\|W_n - f\|_{\infty} \stackrel{p}{\to} 0$  as  $n \to \infty$ , then  $\int_0^1 W_n(t) dt \stackrel{p}{\to} \int_0^1 f(t) dt$ . Is this conjecture true or false? If true, give a proof; if false, give a counterexample.

### 2. Lognormal MLE

If  $Z \sim N(\mu, \sigma^2)$ , then  $X = e^Z$  has the lognormal distribution with parameters  $\mu$  and  $\sigma^2$ . In some situations a threshold  $\gamma$ , included by taking

$$X = \gamma + e^Z$$

may be desirable, and in this case X is said to have the three-parameter lognormal distribution with parameters  $\gamma, \mu$ , and  $\sigma^2$ . Let data  $X_1, \ldots, X_n$  be i.i.d. from this three-parameter lognormal distribution.

- (a) Find the common marginal density for the  $X_i$ .
- (b) Suppose the threshold  $\gamma$  is known. Find the maximum likelihood estimators  $\hat{\mu} = \hat{\mu}(\gamma)$  and  $\hat{\sigma}^2 = \hat{\sigma}^2(\gamma)$  of  $\mu$  and  $\sigma^2$ . (Assume  $\gamma < X_{(1)}$ .)
- (c) Let  $\ell(\gamma, \mu, \sigma^2)$  denote the log-likelihood function. The maximum likelihood estimator for  $\gamma$ , if it exists, will maximize  $\ell(\gamma, \hat{\mu}(\gamma), \hat{\sigma}^2(\gamma))$  over  $\gamma$ . Determine

$$\lim_{\gamma \uparrow X_{(1)}} \ell(\gamma, \hat{\mu}(\gamma), \hat{\sigma}^2(\gamma)).$$

Hint: Show first that as  $\gamma \uparrow X_{(1)}$ ,

$$\hat{\mu}(\gamma) \sim \frac{1}{n} \log(X_{(1)} - \gamma), \quad \text{and } \hat{\sigma}^2(\gamma) \sim \frac{n-1}{n^2} \log^2(X_{(1)} - \gamma),$$

where the notation  $f(\gamma) \sim g(\gamma)$  means  $f(\gamma)/g(\gamma) \to 1$ .

## 3. Censored Poisson

Suppose that  $Y_i$ , i = 1, ..., n are drawn i.i.d. from a  $Pois(\theta)$  distribution for  $\theta \in (0, \infty)$ , but that we only observe the censored quantities

$$X_i = \begin{cases} 0 & \text{if } Y_i = 0\\ 1 & \text{if } Y_i > 0. \end{cases}$$

(a) Under what conditions does an MLE  $\hat{\theta}$  of  $\theta$  based on  $(X_1, \dots, X_n)$  exist? Compute the probability that the  $\hat{\theta}$  exists as a function of  $\theta$  and n. What happens as  $n \to +\infty$ ?

(b) Define the modified estimator

$$\delta(X) = \begin{cases} \hat{\theta}(X) & \text{when the MLE exists} \\ 1 & \text{otherwise.} \end{cases}$$

Prove that  $\sqrt{n}(\delta - \theta) \Rightarrow N(0, e^{\theta} - 1)$ .

4. MLE Under Reparameterization

Given a family  $\widehat{\mathcal{P}} = \{p_{\theta}(x) : \theta \in \Theta\}$  of density functions, let  $\widehat{\theta}$  denote the MLE of  $\theta$ . Suppose h is a one-to-one function from  $\Theta$  to  $h(\Theta)$ . Setting  $\eta = h(\theta)$ , and  $q_{\eta}(x) = p_{h^{-1}(\eta)}(x)$ , define the family  $\mathcal{Q} = \{q_{\eta}(x) : \eta \in h(\Theta)\}$ , meaning that we have reparameterized the model using  $\eta$ .

- (a) Show that the MLE  $\hat{\eta}$  of  $\eta$  is  $h(\hat{\theta})$ . (This result shows that MLEs are invariant under one-to-one reparameterizations.)
- (b) If  $\hat{\theta}$  is consistent for  $\theta$ , then must  $\hat{\eta}$  be consistent for  $\eta$ ? Prove or give a counterexample.

5. Superefficiency Revisited

Recall the superefficient estimator for  $\theta$  given  $X_1, \ldots, X_n \overset{\text{i.i.d.}}{\sim} N(\theta, 1)$ , from the last problem set:

$$\delta_n = \begin{cases} \bar{X}_n & |\bar{X}_n| > n^{-1/4} \\ 0 & |\bar{X}_n| \le n^{-1/4} \end{cases}, \text{ where } \bar{X}_n = \frac{1}{n} \sum_{i=1}^n X_i.$$

Show that, pointwise in  $\theta$ , as  $n \to \infty$ ,

$$n \operatorname{MSE}(\delta_n; \theta) \to 1\{\theta \neq 0\},$$

but that the convergence is not uniform in  $\theta$ ; in fact,

$$\sup_{\theta \in \mathbb{R}} |n \operatorname{MSE}(\delta_n; \theta) - 1\{\theta \neq 0\}| \to \infty$$

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(Hint: consider the sequence  $\theta_n = n^{-1/4}$ )