

## Statistics 581, Problem Set 9

Wellner; 11/27/2002

**Reading:** Chapter 4, Sections 1-2;

Ferguson, ACLST, Chapter 20, pages 133-139; Chapter 22, pages 144-150;

Lehmann and Casella, Chapter 6, especially section 6.5, pages 461-468.

**Due:** Wednesday, December 4, 2002.

- (a) Lehmann and Casella, problem 6.3.1, page 501.  
(b) Lehmann and Casella, problem 6.3.2, page 501.  
(c) Lehmann and Casella, problem 6.3.4, page 501.  
(d) Lehmann and Casella, problem 6.3.18, page 502. [**Note:** It seems to me that 3.15(b) should be 3.15(c) since  $C(0, a)$  is a *scale family*.]
- Suppose that  $(Y|Z) \sim \text{Weibull}(\lambda^{-1}e^{-\gamma Z}, \beta)$ , as in problem 2 of problem set # 8, where  $Z \sim \text{Bernoulli}(\eta)$ . Reparametrize as in problem set # 8 so that the joint density is  $p_\theta(y, z)$  with  $\theta = (\theta_1, \theta_2, \theta_3, \theta_4)$  where  $\theta_4 = \eta$ . Thus the conditional hazard function is

$$\lambda_\theta(t|z) = \theta_3\theta_2 \exp(\theta_1 z) t^{\theta_3 - 1},$$

$Z \sim \text{Bernoulli}(\eta)$ , and  $\theta = (\theta_1, \theta_2, \theta_3, \theta_4)$  with  $\theta_4 \equiv \eta$ . Let  $X = (Y, Z)$ , and suppose that we observe  $X_1, \dots, X_n$  i.i.d. as  $X$ .

- Find the score equations for estimation of  $\theta$ .
  - Give conditions on the data  $X_1, \dots, X_n = (Y_1, Z_1), \dots, (Y_n, Z_n)$  guaranteeing that the score equations have a unique solution which maximizes the likelihood. Call the resulting estimators  $\hat{\theta}_n$ .
  - What does theorem 4.1.2 (Chapter 4, page 5), say about the asymptotic distribution of  $\sqrt{n}(\hat{\theta}_n - \theta_0)$  when the distribution of the data is given by  $P_{\theta_0}$ .
  - Suppose that  $\theta_1 \neq \theta_0$  is the “true” value of the parameter  $\theta$ , and we consider the likelihood ratio  $L_n(\theta_1)/L_n(\theta_0)$  where  $L_n(\theta) \equiv \prod_{i=1}^n p_\theta(X_i)$ . Show that  $n^{-1} \log(L_n(\theta_1)/L_n(\theta_0)) \rightarrow_p$  some constant, and identify the constant as explicitly as possible in terms of  $\theta_1, \theta_0$ .
- Consider the Weibull family of example 3.2.5:  $\mathcal{P} = \{P_\theta : \theta \in \Theta\}$  with  $\Theta \subset R^{+2}$  given by the (Lebesgue) densities

$$p_\theta(x) = \frac{\beta}{\alpha} \left(\frac{x}{\alpha}\right)^{\beta-1} \exp\left(-\left(\frac{x}{\alpha}\right)^\beta\right) 1_{[0, \infty)}(x)$$

where  $\theta \equiv (\alpha, \beta) \in (0, \infty) \times (0, \infty) \subset R^2$ . Suppose that  $X, X_1, \dots, X_n$  are i.i.d. with density function  $p_\theta$ .

A. If  $X \sim P_\theta \in \mathcal{P}$ , show that the distributions of  $\log X$  form a location and scale family from a Gumbel (extreme value) density on  $R$ .

B. Use the result of A to construct method of moments estimators or quantile based estimators  $\bar{\theta}_n$  of  $\theta = (\alpha, \beta)$ .

C. Show that the method of moments or quantile estimators  $\bar{\theta}_n$  of  $\theta$  are asymptotically normal, and find the asymptotic distribution; i.e. show that

$$\sqrt{n}(\bar{\theta}_n - \theta) \rightarrow_d N_2(0, \Sigma) \quad \text{for some} \quad \Sigma.$$

- D. Does a maximum likelihood estimate of  $\hat{\theta} = (\hat{\alpha}, \hat{\beta})$  exist? Is it unique?  
 E. Compute an approximate (one - step) maximum likelihood estimate  $\tilde{\theta}$  of  $\theta$  using the method of moment estimators  $\bar{\theta}_n$  as the preliminary estimators based on the following data (with  $n = 19$ ):

0.19, 0.78, 0.96, 1.31, 2.78, 3.16, 4.15, 4.67, 4.85, 6.50,  
 7.35, 8.01, 8.27, 12.06, 31.75, 32.52, 33.91, 36.71, 72.89 .

[These are failure times in minutes for an insulating fluid between two electrodes subject to a voltage of 34 kV. – from Nelson, *Applied Life Data Analysis*, page 105.]

- F. Compute the maximum likelihood estimator  $\hat{\theta}_n$ , and compare it with the one step estimator computed in E.

4. **Optional bonus problem:** Lehmann and Casella, TPE, problem 6.3.22, page 503, reworded as follows. (In other words, prove (vi) of theorem 1.5, page 5, chapter 4 notes). Suppose that  $X_1, \dots, X_n$  are i.i.d. with density  $p_\theta$ ,  $\theta \in \Theta \subset R^k$ , satisfying the hypotheses of theorem 4.1, page 429 (the Cramér conditions given in (A) - (D) on page 429). Show that the following Local Asymptotic Normality (LAN) result holds for the (local) log- likelihood ratios: with

$$L_n(\theta) \equiv \log\left(\prod_{i=1}^n p_\theta(X_i)\right) = \sum_{i=1}^n \log p_\theta(X_i),$$

for a fixed  $\theta_0 \in \Theta$ ,

$$\begin{aligned} L_n(\theta_0 + n^{-1/2}\underline{t}) - L_n(\theta_0) &= \frac{1}{\sqrt{n}} \sum_{i=1}^n \underline{t}^T \dot{l}_\theta(X_i) - \frac{1}{2} \underline{t}^T I(\theta_0) \underline{t} + o_p(1) \\ &\rightarrow_d N(0, \underline{t}^T I(\theta_0) \underline{t}) - \frac{1}{2} \underline{t}^T I(\theta_0) \underline{t} =_d N\left(-\frac{1}{2} \sigma^2 \sigma^2\right) \end{aligned}$$

under  $P_{\theta_0}$  where  $\sigma^2 \equiv \underline{t}^T I(\theta_0) \underline{t}$ .