# UC Berkeley Department of Statistics

## STAT 210A: Introduction to Mathematical Statistics

## Problem Set 8

Fall 2014

# Problem 8.1

Consider estimating success probabilities  $\theta_1, \ldots, \theta_p$  for p independent binomial variables,  $X_1, \ldots, X_p$ , each based on m trials, under compound squared error loss,  $L(\theta, d) = \sum_{i=1}^{p} (\theta_i - d_i)^2$ .

- (a) Following a Bayesian approach, model the unknown parameters as random variables  $\Theta_1, \ldots, \Theta_p$  that are i.i.d. from a beta distribution with parameters  $\alpha$  and  $\beta$ . Determine the Bayes estimators of  $\Theta_1, \ldots, \Theta_p$ .
- (b) In the Bayesian model,  $X_1, \ldots, X_p$  are i.i.d. Determine the first two moments for their common marginal distribution,  $EX_i$  and  $EX_i^2$ . Using these, suggest simple method of moments estimators for  $\alpha$  and  $\beta$ .
- (c) Give empirical Bayes estimators for  $\theta_i$  combining the simple "empirical" estimates for  $\alpha$  and  $\beta$  in (b) with the Bayes estimate for  $\theta_i$  when  $\alpha$  and  $\beta$  are known in (a).

## Problem 8.2

Let  $X_1, \ldots, X_n$  be an i.i.d. sample from the uniform distribution on  $[0, \theta]$ .

- (a) Consider the problem of testing  $H_0: \theta \leq \theta_0$  versus  $H_1: \theta > \theta_0$ . Show that any test  $\delta$  for which  $\mathbb{E}_{\theta_0}[\delta(X)] = \alpha$ ,  $\mathbb{E}_{\theta}[\delta(X)] \leq \alpha$  for all  $\theta \leq \theta_0$  and  $\delta(x) = 1$  when  $x_{(n)} = \max\{x_1, \ldots, x_n\} > \theta_0$  is UMP at level  $\alpha$ .
- (b) Now consider the problem of testing  $H_0: \theta = \theta_0$  against  $H_1: \theta \neq \theta_0$ . Show that a unique UMP test exists, and is given by

$$\delta(x) = \begin{cases} 1 & \text{if } x_{(n)} > \theta_0 \text{ or } x_{(n)} < \theta_0(\alpha)^{1/n} \\ 0 & \text{otherwise.} \end{cases}$$

## Problem 8.3

Suppose that  $X_1, ..., X_n$  are independent exponential random variables with  $\mathbb{E}[X_i] = \beta t_i$  where  $t_1, ..., t_n$  are known constants, and  $\beta > 0$  is an unknown parameter.

1

- (a) Show that the MLE of  $\beta$  is given by  $\widehat{\beta}_n = \frac{1}{n} \sum_{i=1}^n X_i/t_i$ .
- (b) Prove that  $\sqrt{n} \left( \beta \widehat{\beta}_n \right) \stackrel{d}{\to} N(0, \beta^2)$ .

(c) Suppose that we want to test  $H_0: \beta = 1$  versus  $H_1: \beta \neq 1$ . In order to do so, we consider the statistic

$$G(X_1, \dots, X_n) = \log p(X; \widehat{\beta}_n) - \log p(X; 1),$$

where  $p(X; \beta) = \prod_{i=1}^{n} p(X_i; \beta, t_i)$  is the likelihood. Show that  $G(X) = n \left( \widehat{\beta}_n - \log \widehat{\beta}_n - 1 \right)$ .

(d) Show that when  $H_0$  is true, we have  $2G(X) \stackrel{d}{\rightarrow} \chi_1^2$ .

## Problem 8.4

Consider simple versus simple testing from a Bayesian perspective. Let  $\Theta$  have a Bernoulli distribution with  $P(\Theta = 1) = p$  and  $P(\Theta = 0) = 1 - p$ . Given  $\Theta = 0$ , X will have density  $p_0$  and given  $\Theta = 1$ , X will have density  $p_1$ .

(a) Show that the chance of accepting the wrong hypothesis in the Bayesian model using a test function  $\varphi$  is

$$R(\varphi) = E[I(\Theta = 0)\varphi(X) + I(\Theta = 1)(1 - \varphi(X))].$$

- (b) Use the tower property to relate  $R(\varphi)$  to  $E_0(\varphi) = E(\varphi(X)\gamma\Theta = 0)$  and  $E_1(\varphi) = E(\varphi(X)\gamma\Theta = 1)$ .
- (b) Find the test function  $\varphi^*$  minimizing  $R(\varphi)$ . Show that  $\varphi^*$  is a likelihood ratio test, identifying the critical value k.