

Stat 643 Exam 2

May 3, 2007

**I have neither given nor received unauthorized assistance on this examination.**

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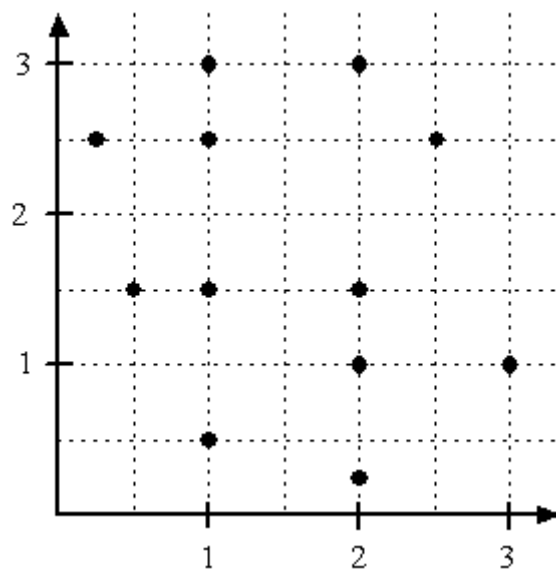
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There are 10 "small" problems on this exam. Some are easier than others. All will be scored out of 10 points to make a total possible score of 100 points.

1. Suppose that the figure below shows risk vectors for all non-randomized rules in a 2-state decision problem. Identify (give the coordinates of) the (possibly randomized) risk vectors that are
- minimax
  - Bayes versus a prior that is uniform on  $\Theta = \{\theta_1, \theta_2\}$



2. Suppose that  $X$  and  $Y$  are independent,  $X \sim \text{Bi}(n, p)$  and  $Y \sim \text{Ber}(p)$ . Consider squared error loss estimation of  $p$  and an estimator  $\hat{p} = \delta(X)$ . Let  $W = X + Y$ . Find an estimator  $\hat{p}^* = \gamma(W)$  that improves upon  $\hat{p}$ . (Carefully say why your new estimator is better in terms of risk.)

3. Consider a model for an observable  $X$  with real parameter  $\alpha \in (0,1)$  and pdf on  $(0,1)$

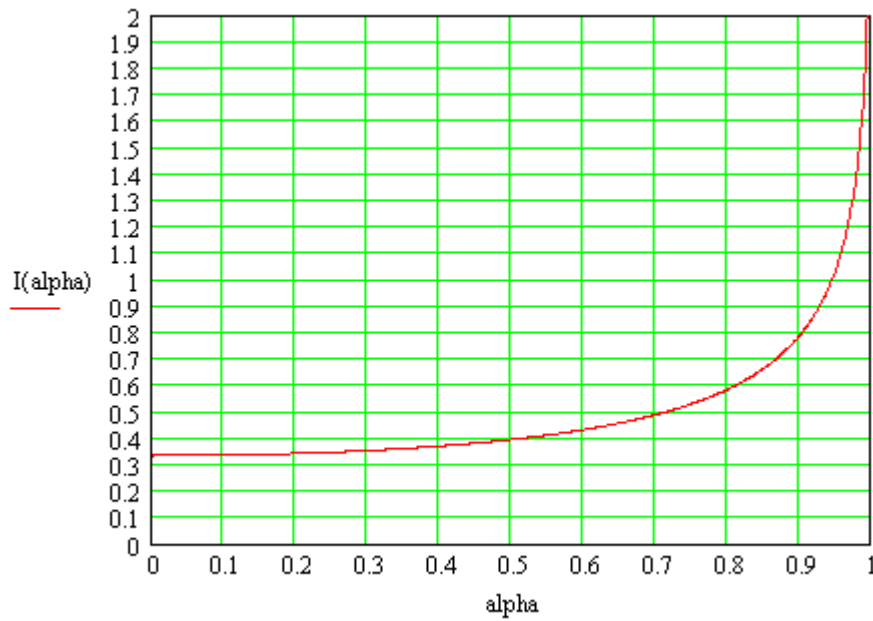
$$f(x|\alpha) = 1 + \alpha(2x - 1)$$

Write out explicitly as possible the Fisher information in  $X$  about  $\alpha$  at  $\alpha_0$ ,  $I_X(\alpha_0)$ . (You'll probably have to leave this in terms of an integral that would need to be evaluated numerically.)

4. In the context of problem 3,  $I_x(\alpha)$  can be computed numerically and looks as in the plot below.

Based on the information in this plot, if  $X_1, X_2, \dots, X_{100}$  are iid with marginal density  $f(x|\alpha)$ ,

- what do you propose as an approximation to the distribution of an "MLE" of  $\alpha$  if it is the case that  $\alpha = .5$ , and
- what do you propose as (realizable/implementable) "large sample conservative" approximately 95% two-sided confidence limits for  $\alpha$ , based on an "MLE" of  $\alpha$ , say  $\hat{\alpha}_{100}$ ?



5. Again in the context of problem 3, if  $X_1, X_2, \dots, X_n$  are iid with marginal density  $f(x|\alpha)$ ,
- what is the likelihood equation (that would need to be solved in order to find an MLE of  $\alpha$ )?
  - give an *explicit* formula for an estimator of  $\alpha$  that will be "asymptotically efficient" here.

6. Consider the problem of likelihood ratio testing in the *non-regular* family of Uniform( $0, \theta$ ) distributions. That is, suppose that  $X_1, X_2, \dots, X_n$  are iid Uniform( $0, \theta$ ) and consider testing  $H_0: \theta = \theta_0$  versus  $H_a: \theta \neq \theta_0$ . Let

$$\Lambda_n = 2 \left( \sup_{\theta} L_n(\theta) - L_n(\theta_0) \right)$$

where (as usual)  $L_n(\theta)$  is the  $n$  observation log-likelihood. What is the large sample distribution of  $\Lambda_n$  under the null hypothesis? (Hint: You may use without proof the facts that the MLE of  $\theta$  is  $\max_{i=1,2,\dots,n} X_i$ , and that under the null hypothesis,  $n \left( \theta_0 - \max_{i=1,2,\dots,n} X_i \right)$  is asymptotically Exponential with mean  $\theta_0$ .)

7. Suppose that as on the Exponential Families handout, for  $\eta \in \Gamma \subset \mathfrak{R}$ , distributions  $P_\eta$  have densities

$$f_\eta(x) = K(\eta) \exp(\eta T(x)) h(x)$$

If  $X_1, X_2, \dots, X_n$  are iid  $P_{\eta_0}$  and  $L_n(\eta)$  is the  $n$  observation log-likelihood, to what does  $\frac{1}{n} L_n(\eta)$  converge in probability? (Give a formula in terms of the factors of  $f_\eta$ .)

8. Bayesians sometimes argue that their "posteriors are consistent." Consider the simplest possible version of this. Suppose that  $\Theta = \{1, 2\}$  and for some  $\sigma$ -finite measure  $\mu$ ,  $f_1$  and  $f_2$  are densities for two different probability distributions  $P_1$  and  $P_2$ . For a prior distribution  $G$  on  $\Theta$ , with  $g_i = G(\{i\}) \in (0, 1)$  for  $i = 1, 2$  and  $X_1, X_2, \dots, X_n$  iid with marginal one of  $P_1$  and  $P_2$ ,

- a) what is the posterior probability (based on the  $n$  observations) that  $\theta = 1$ ?
- b) argue carefully that under the  $\theta = 1$  model, the posterior probability that  $\theta = 1$  from a) converges to 1 in probability.

9. Suppose that  $X \sim \text{Bi}(2, p)$  and that squared error loss of  $p \in [0, 1]$  is under consideration. Argue directly that for any  $c \in (0, 1)$  the estimator

$$\delta_c(X) = \begin{cases} 0 & \text{if } x = 0 \\ c & \text{if } x = 1 \\ 1 & \text{if } x = 2 \end{cases}$$

is admissible. (Hint: What must be the  $p = 0$  risk and the  $p = 1$  risk of any  $\phi$  improving upon a  $\delta_c$ . What does that say about the form of  $\phi$ ?)

10. An alternative to 0-1 loss for decision-theoretic treatments of some testing problems is the so-called "linear loss." That is, for an interval  $\Theta \subset \mathfrak{R}$  and action space  $\mathcal{A} = \{0,1\}$ , if  $\theta_0 \in \Theta$ , testing  $H_0: \theta \leq \theta_0$  vs  $H_a: \theta > \theta_0$  might be phrased in decision theoretic terms using a loss function

$$L(\theta, a) = \begin{cases} \max\{0, \theta - \theta_0\} & \text{for } a = 0 \\ \max\{0, \theta_0 - \theta\} & \text{for } a = 1 \end{cases}$$

For a prior  $G$  on  $\Theta$ , and distributions  $P_\theta$  of an observable  $X$ , what is the form of a Bayes rule for  $G$ ? (Hint: Consider  $L(\theta, 0) - L(\theta, 1)$ .)