

**MATH 562-01 MATHEMATICAL STATISTICS**

Exam 2 (10/24/16, Monday)

Name: \_\_\_\_\_

1. (10 points) Let  $X_1, X_2, \dots, X_n$  be a random sample from the discrete distribution

$$P[X = x] = \begin{cases} \frac{\theta^{2x} e^{-\theta^2}}{x!} & x = 0, 1, 2, \dots \\ 0 & \text{otherwise} \end{cases}$$

where  $\theta > 0$ . Are the MME  $\hat{\theta}$  and MLE  $\tilde{\theta}$  unbiased? Carefully justify your answers.

2. (10 points each) Let  $X$  and  $Y$  be independent random variables with  $E(X) = 1$ ,  $E(Y) = 2$  and  $\text{var}(X) = \text{var}(Y) = \sigma^2$ . Find the constant  $k$  such that  $T = k(X^2 - Y^2) + Y^2$  is an unbiased estimator for  $\sigma^2$ .
3. (10 points) A sequence of independent Bernoulli trials with probability of success  $p$  is performed. Let  $X$  denote the number of trials until the first success. Four independent realizations of  $X$  are obtained:  $x_1 = 1, x_2 = 1, x_3 = 3$ , and  $x_4 = 1$ . If, a priori,  $p$  has probability density function  $h(p) = 2p$ , for  $p \in (0, 1)$ , and squared errors loss  $L(t, p) = (t - p)^2$  is used, find the Bayesian estimate for  $p$ .
4. (20 points) A random sample of size  $n$  is drawn from a normal population  $N(\mu_1, 2)$ , and another random sample of the same size is drawn independently from another normal population  $N(\mu_2, 5)$ .

(1). Find the MLE  $\tilde{\theta}$  for  $\theta = \mu_1 - \mu_2$ . (NOTE: Specify the distribution you use, and justify every step carefully.)

(2). Show that  $\tilde{\theta}$  is a UMVUE, so it is an efficient estimator.

5. (20 points) Let  $X_1, X_2, \dots, X_n$  be a random sample from  $UNIF(0, \theta)$ . We know that the MLE  $\tilde{\theta} = \max\{X_1, X_2, \dots, X_n\}$  is biased. Also, the MLE  $\tilde{\theta}$  has probability density function

$$f_{\tilde{\theta}}(x) = \frac{nx^{n-1}}{\theta^n}, \text{ for } x \in (0, \theta).$$

(1). Find the constant  $k$  such that  $T = k\tilde{\theta}$  is an unbiased estimator. Is  $\tilde{\theta}$  itself asymptotically unbiased?

(2). Find  $\text{var}(T)$ , the variance of  $T$ , and  $\frac{1}{E\left[\left(\frac{\partial}{\partial \theta} \ln L(\theta)\right)^2\right]}$ , the lower bound given in the

Cramer-Rao Theorem. Compare them and explain why the CRLB does not hold in this case.

6. (10 points) Let  $X_1, X_2, \dots, X_n$  be a random sample from  $BIN(m, p)$ . Show that the MLE  $\tilde{p} = \frac{\bar{X}}{m}$  for  $p$  is efficient, i.e., a UMVUE. (Hint: You may use known results for  $BIN(m, p)$ .)

7. (20 points) Let  $X_1, X_2, \dots, X_n$  be a random sample from an exponential population  $EXP(\theta)$ , where  $\theta \in (0, \infty)$ , and  $T = u(X_1) = \begin{cases} 0, & 0 < X_1 < 1 \\ 1, & X_1 \geq 1 \end{cases}$ .

(1). Show that  $T$  is an unbiased estimator for  $P[X > 1] = e^{-\frac{1}{\theta}}$ .

(2). By the Rao-Blackwell Theorem,  $S = E\left[T \middle| \sum_{i=1}^n X_i\right]$  is an efficient estimator

of  $P[X > 1] = e^{-\frac{1}{\theta}}$ . Find  $S$ . (Hint: Use  $f_{X_1 \mid \sum_{i=1}^n X_i = s}(x_1) = \frac{n-1}{s^{n-1}}(s - x_1)^{n-2}$ ,  $x_1 \in (1, s)$ .)

NOTE: We just apply the Rao-Blackwell Theorem here, without proof. Rao-Blackwell provides a way to find efficient estimator by iteration, a quite effective method.