## GROUND RULES:

- This exam contains 6 questions. The questions are of equal weight.
- Print your name at the top of this page in the upper right hand corner.
- This exam is closed book and closed notes.
- Show all of your work and explain all of your reasoning! **Translation:** No work, no credit. Insufficient explanation, no credit. If you are unsure about whether or not you should explain a result or step in your derivation/proof, then this means you probably should explain it.
- Do not talk with anyone else about this exam. You must work by yourself. No communication of any type with others.

1. Suppose  $X_1, X_2, ..., X_n$  is an iid sample from

$$f_X(x|\theta) = \begin{cases} e^{-(x-\theta)}, & x \ge \theta\\ 0, & \text{otherwise}, \end{cases}$$

where  $-\infty < \theta < \infty$  is unknown.

(a) The method of moments estimator of  $\theta$  is  $\hat{\theta} = \overline{X} - 1$ . Use the Central Limit Theorem and Delta Method to derive the asymptotic distribution of  $\sqrt{n}(\hat{\theta} - \theta)$ .

(b) The maximum likelihood estimator of  $\theta$  is  $X_{(1)}$ , the minimum order statistic. Find a sequence of constants  $a_n$  such that  $a_n(X_{(1)} - \theta)$  converges in distribution (a non-degenerate one) and state the asymptotic distribution.

(c) Use the asymptotic distributions to approximate  $var(\overline{X}-1)$  and  $var(X_{(1)})$  for large n. Which one is smaller?

- 2. Suppose that  $X_1, X_2, ..., X_n$  is an iid sample from a  $\mathcal{N}(0, \sigma^2)$  distribution, where  $\sigma^2 > 0$ .
- (a) Show that  $U = U(\mathbf{X}) = \sum_{i=1}^{n} X_i^2$  is a sufficient statistic.

(b) Show directly that  $T = T(\mathbf{X}) = \sum_{i=1}^{n} X_i$  is not sufficient by showing that  $f_{\mathbf{X}|T}(\mathbf{x}|t)$  is not free of  $\sigma^2$ .

(c) Define  $\hat{\sigma}^2 = \frac{1}{n} \sum_{i=1}^n X_i^2$ . Show that

$$\sqrt{n}(\widehat{\sigma}^2 - \sigma^2) \xrightarrow{d} \mathcal{N}(0, 2\sigma^4).$$

(d) Use the result in part (c), along with Slutsky's Theorem, to show that

$$Z_n = \sqrt{\frac{n}{2}} \left( \frac{\widehat{\sigma}^2 - \sigma^2}{\widehat{\sigma}^2} \right) \stackrel{d}{\longrightarrow} \mathcal{N}(0, 1).$$

3. Suppose that  $X_1, X_2, ..., X_n$  is an iid sample from

$$f_X(x|\alpha,\beta) = \alpha \beta^{\alpha} \left(\frac{1}{x+\beta}\right)^{\alpha+1} I(x>0),$$

where  $\alpha > 0$  and  $\beta > 0$ .

(a) Show that  $(X_{(1)}, X_{(2)}, ..., X_{(n)})$  is a sufficient statistic for the two-parameter family  $\{f_X(x|\alpha, \beta) : \alpha > 0, \beta > 0\}$  and that no further reduction is possible.

(b) Consider the one-parameter subfamily  $\{f_X(x|\alpha) : \alpha > 0\}$  arising when  $\beta = \beta_0$  is known. Find a one-dimensional sufficient statistic for this subfamily.

(c) Is your sufficient statistic in part (b) complete for the subfamily  $\{f_X(x|\alpha) : \alpha > 0\}$ ? Explain.

4. The discrete random vector (X, Y) has a joint probability mass function  $f_{X,Y}(x, y)$  described in the following table:

$$\begin{array}{ccc} y = 1 & y = 2 \\ \hline x = 1 & f_{X,Y}(1,1) = \frac{1}{4} - \theta & f_{X,Y}(1,2) = \frac{1}{4} + \theta \\ x = 2 & f_{X,Y}(2,1) = \frac{1}{4} + \theta & f_{X,Y}(2,2) = \frac{1}{4} - \theta \end{array}$$

or, written more succinctly,

$$y = 1 \qquad y = 2$$
$$x = 1 \quad \frac{1}{4} - \theta \qquad \frac{1}{4} + \theta$$
$$x = 2 \quad \frac{1}{4} + \theta \qquad \frac{1}{4} - \theta$$

Note that the support of (X, Y) is  $\mathcal{X} = \{(1, 1), (1, 2), (2, 1), (2, 2)\}$ . The parameter space is  $\Theta = \{\theta : -\frac{1}{4} < \theta < \frac{1}{4}\}$ . We observe a single (X, Y).

(a) Show that both X and Y are ancillary statistics.

- (b) Show that U = X + Y is not ancillary.
- (c) Show that V = |X Y| is a sufficient statistic for  $\theta$ .
- (d) Show that V is a complete statistic by using the definition; that is, show that

$$E_{\theta}[g(V)] = 0 \quad \forall \theta \in \Theta \implies P_{\theta}(g(V) = 0) = 1 \quad \forall \theta \in \Theta.$$

5. Suppose that  $X_1, X_2, ..., X_n$  is an iid sample from

$$f_X(x|\theta) = \begin{cases} \theta^2 x e^{-\theta x}, & x > 0\\ 0, & \text{otherwise}, \end{cases}$$

where  $\theta > 0$ .

(a) Show that  $\{f_X(x|\theta) : \theta > 0\}$  is a scale family with scale parameter  $1/\theta$ . What is the standard density?

(b) Let  $\overline{X} = n^{-1} \sum_{i=1}^{n} X_i$  denote the sample mean. Show that  $\overline{X}$  is a complete and sufficient statistic.

(c) Let  $X_{(n)}$  denote the maximum order statistic. Show that

$$E(X_{(n)}|\overline{X}) = \frac{\overline{X}E(X_{(n)})}{E(\overline{X})}.$$

6. Suppose that  $X_1, X_2, ..., X_n$  is an iid sample from a lognormal distribution; i.e., the population probability density function is

$$f_X(x|\mu,\sigma^2) = \frac{1}{\sqrt{2\pi\sigma}x} e^{-(\ln x - \mu)^2/2\sigma^2} I(x > 0),$$

where  $-\infty < \mu < \infty$  and  $\sigma^2 > 0$ .

(a) Derive E(X) and  $E(X^2)$ . Then find the method of moments estimators of  $\mu$  and  $\sigma^2$ . *Hint:* Recall that if  $U \sim \mathcal{N}(\mu, \sigma^2)$ , then  $X = e^U$  is lognormal.

(b) Find the maximum likelihood estimators of  $\mu$  and  $\sigma^2$ . Don't worry about verifying 2nd order conditions.