



Note: please fully justify all your answers.

1. Let  $X_1, \dots, X_n$  denote a random sample of size  $n$  from an  $\text{Ex}(\theta)$  population, where  $\theta > 0$ . Let  $X_{(1)} = \min\{X_1, \dots, X_n\}$  and  $\bar{X} = \sum_{i=1}^n X_i/n$ . In this context, show that, as  $n \rightarrow +\infty$ ,

a)  $a_n X_{(1)} \xrightarrow{d} \text{Ex}(\theta)$  as long as the sequence of real numbers  $(a_n)$  is such that  $\lim_{n \rightarrow +\infty} n/a_n = 1$ . (1.5)

b)  $b_n X_{(1)} \xrightarrow{P} 0$  as long as the sequence of real numbers  $(b_n)$  is such that  $\lim_{n \rightarrow +\infty} n/b_n = +\infty$ . (1.5)

c)  $nX_{(1)}(1 - X_{(1)}) \xrightarrow{d} \text{Ex}(\theta)$ . (2.0)

d)  $2\sqrt{n}(\sqrt{\theta\bar{X}} - 1) \xrightarrow{d} N(0, 1)$ . (2.0)

2. Consider a bivariate population  $(X, Y)$  whose joint probability distribution is such that, for some  $\lambda > 0$ ,

$$X \mid \lambda \sim \text{Ex}(1/\lambda) \quad \text{and} \quad Y \mid X = x, \lambda \sim N(0, \lambda x) .$$

Denote by  $(X_1, Y_1), \dots, (X_n, Y_n)$  a random sample of size  $n$  extracted from this population. In this context,

a) Show that  $T = \sum_{i=1}^n X_i + \frac{1}{2} \sum_{i=1}^n Y_i^2/X_i$  is a complete and sufficient statistic for  $\lambda$ . (2.0)

b) Verify that  $E[Y^2/X \mid \lambda] = \lambda$ . (1.5)

c) Show that  $S = 2T/(3n)$  is the UMVU estimator of  $\lambda$ . (1.5)

d) Is  $S$  also the most efficient estimator of  $\lambda$ ? (1.5)

e) Determine  $\text{Var}(S \mid \lambda)$ . (1.5)

3. Let  $X_1, \dots, X_n$  denote a random sample of size  $n$  from a population  $X$  with probability density function given by

$$f(x \mid \theta) = \frac{2}{\theta^2} x, \quad 0 < x < \theta$$

where  $\theta > 0$ . Denote by  $x_1, \dots, x_n$  the corresponding observed sample. The goal is to implement a Bayesian analysis of these data.

- a) Suppose that, *a priori*,

$$\pi(\theta) \propto \frac{1}{\theta^2} I_{(1, +\infty)}(\theta) .$$

Does this *a priori* distribution correspond to a proper probability distribution? If so, determine the corresponding normalizing constant. (1.5)

- b) Assuming the *a priori* distribution described above, show that

$$\pi(\theta \mid x_1, \dots, x_n) = (2n + 1) (x_{(n)} \vee 1)^{2n+1} \theta^{-(2n+2)} I_{(x_{(n)} \vee 1, +\infty)}(\theta)$$

where  $x_{(n)}$  denotes the sample maximum and  $a \vee b \equiv \max\{a, b\}$ . (2.0)

- c) Before observing  $X_{n+1}$ , you are given the possibility of betting on the realization of two events:  $\{X_{n+1} \leq 1\}$  or  $\{X_{n+1} > 1\}$ . Taking into account the Bayesian analysis herein developed, devise a strategy that, as a function of  $x_1, \dots, x_n$ , tells you on which of the events one should bet. (1.5)