## Endterm AM2080 2022-2023

13:30 - 16:30 November 11, 2022

This written endterm exam contains 5 questions, each question counts for 20% of the final grade of the written test. You are only allowed to use a personally made cheat-sheet, the sheet with information on probability distributions, and the tables for normal, binomial, chi-square and Student-t distributions. You are not allowed to use any books or notes.

1. Let  $X_1, \ldots, X_n$  be independent random variables with (marginal) probability density

$$p_{\theta}(x) = \frac{1}{2}\theta e^{-\theta|x|}, \text{ for } x \in (-\infty, \infty),$$

where  $\theta > 0$  is an unknown parameter.

- (a) Determine the maximum likelihood estimator for  $\theta^2$ .
- (b) Determine the Bayes estimator for  $\theta^2$ , with respect to the prior density given by  $\pi(\theta) = e^{-\theta}$  for  $\theta > 0$  and 0 elsewhere.
- 2. Let  $X_1, \ldots, X_n$  be a sample from a  $N(\mu, 4)$  distribution. We test  $H_0: \mu \geq 1$  against  $H_1: \mu < 1$  at level  $\alpha_0 = 0.05$  with test statistic  $T = \sqrt{n}(\overline{X} 1)/2$ .
  - (a) Show that the critical region of the test is given by

$$K = \left\{ (x_1, \dots, x_n) : \sqrt{n} \frac{\overline{x} - 1}{2} \le -1.645 \right\}.$$

- (b) Show that the power function is given by  $\pi(\mu; K) = \Phi(-1.645 + \sqrt{n}(1-\mu)/2)$ .
- (c) Determine how large n must be, such that the probability of a type II error in  $\mu=0$  is at most 0.1.
- 3. Let  $X_1, \ldots, X_n$  be independent identically distributed random variables with a probability density

$$p_{\theta}(x) = \begin{cases} \frac{2x}{\theta^2} & 0 \le x \le \theta; \\ 0 & \text{otherwise,} \end{cases}$$

where  $\theta > 0$  is unknown.

(a) Show that  $X_{(n)}/\theta$  is a pivot and that it has distribution function

$$P_{\theta}\left(\frac{X_{(n)}}{\theta} \le t\right) = \begin{cases} 0 & t < 0; \\ t^{2n} & 0 \le t \le 1; \\ 1 & t > 1. \end{cases}$$

- (b) Using the pivot from part (a), find d > 1 such that  $[X_{(n)}, dX_{(n)}]$  is a confidence interval for  $\theta$  with confidence level  $1 \alpha$ .
- (c) We want to test  $H_0: \theta = 1$  against  $H_1: \theta \neq 1$  at significance level  $\alpha$  with test statistic  $X_{(n)}$ . Use part (b) to construct a critical region for this test.

4. Let  $X_1, \ldots, X_n$  be independent random variables with (marginal) probability density

$$p_{\theta}(x) = \frac{\sqrt{2}}{\theta\sqrt{\pi}} \exp\left(-\frac{x^2}{2\theta^2}\right), \quad \text{for } x > 0,$$

where  $\theta > 0$  is an unknown parameter. You may use that  $E_{\theta}[X_1^2] = \theta^2$ .

Show that the maximum likelihood estimator is given by 
$$\hat{\theta} = \sqrt{\frac{1}{n} \sum_{i=1}^{n} X_i^2}$$
.

 $\int$  (b) Determine the likelihood ratio statistic  $\lambda_n$  for testing  $H_0: \theta = \theta_0$  against  $H_1: \theta \neq \theta_0$  and show that

$$\lambda_n = W^{-n} e^{n(W^2 - 1)/2}$$

where  $W = \widehat{\theta}/\theta_0$ .

- $\checkmark$ (c) Compute the Fisher information and give an approximate (two-sided) confidence interval for  $\theta$  with confidence level  $1-\alpha$  based on the asymptotic distribution of the maximum likelihood estimator.
- 5. One tries to fit a suitable linear regression model for observations  $(x_i, y_i)$ , where  $x_i > 0$ , for  $i = 1, \ldots, n$ . Closer inspection of the data reveals a trend in the residuals. In order to take this into account, one suggests the following linear regression model,

$$Y_i = \beta x_i + e_i \sqrt{x_i}, \quad \text{for } i = 1, \dots, n$$

where  $e_1, \ldots, e_n$  are independent such that  $e_i \sim N(0, \sigma^2)$ , and where  $x_1, \ldots, x_n$  are considered to be non-random constants.

Show that  $EY_i = \beta x_i$  and  $var(Y_i) = x_i \sigma^2$ , for i = 1, ..., n. Show that the maximum likelihood estimator for  $\beta$  is given by

$$\widehat{\beta} = \frac{\sum_{i=1}^{n} Y_i}{\sum_{i=1}^{n} x_i}.$$

(a) An alternative estimator is the "average slope" estimator

$$\widetilde{\beta} = \frac{1}{n} \sum_{i=1}^{n} \frac{Y_i}{x_i}.$$

Show that both  $\widehat{\beta}$  and  $\widetilde{\beta}$  are unbiased for  $\beta$ . (d) The mean squared error for  $\hat{\beta}$  is given by

mean squared error for 
$$\beta$$
 is given by

$$MSE(\widehat{\beta}) = \frac{\sigma^2}{\sum_{i=1}^n x_i}.$$

Determine the mean squared error for  $\widetilde{\beta}$ , and show  $MSE(\widehat{\beta}) \leq MSE(\widetilde{\beta})$ . You may use the Cauchy-Schwarz inequality

$$\left(\sum_{i=1}^n a_i b_i\right)^2 \le \left(\sum_{i=1}^n a_i^2\right) \left(\sum_{i=1}^n b_i^2\right)$$