

Exam Introduction to Statistics (AM2080)

November 4, 2024, 13.30–16.30

This written endterm exam contains 5 questions, each question counts for 20% of the final grade of the 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 (other) notes.

1. Let  $X_1, \dots, X_n$  be independent random variables with density

$$p_\theta(x) = \theta x^{-(\theta+1)}, \quad x \geq 1,$$

where  $\theta > 0$ .

- (a) Determine the maximum likelihood estimator for  $\theta$ .  
(b) Determine the Bayes estimator for  $\theta$  based on the prior distribution with density  $\pi(\theta) = e^{-\theta}$  for  $\theta > 0$  and zero elsewhere.
2. Let  $X_1, \dots, X_n \sim \mathcal{N}(\mu, 1)$  and  $Y_1, \dots, Y_m \sim \mathcal{N}(\mu + \delta, 4)$  be independent random variables, with unknown parameters  $\mu, \delta \in \mathbb{R}$ . We want to test the hypothesis  $H_0 : \delta = 0$  against  $H_1 : \delta \neq 0$  via the statistic

$$T_{n,m} = \frac{1}{n} \sum_{i=1}^n X_i - \frac{1}{m} \sum_{j=1}^m Y_j.$$

- (a) Show that  $T_{n,m} \sim \mathcal{N}(0, \frac{m+4n}{nm})$  under  $H_0$ .  
(b) Describe a test procedure to test  $H_0$  against  $H_1$  based on the statistic  $T_{n,m}$  by determining the rejection region for the significance level  $\alpha = 0.1$ .  
(c) Suppose that a test rejects  $H_0$  if  $T_{n,m} > 1$ . Determine the power of this test as a function of  $n, m$ , and  $\delta$ . For  $n = m = 16$ , compute the size of this test, as well as the power against the alternative  $\delta = -1$ .
3. Let  $X_1, X_2, \dots, X_n$  be independent random variables, all with distribution function

$$F_\theta(x) = 1 - e^{-\theta\sqrt{x}} \text{ for } x \geq 0$$

and 0 elsewhere. The parameter  $\theta > 0$  is unknown.

- (a) Show that  $\theta^2 X_{(1)} = \theta^2 \min_{1 \leq i \leq n} X_i$  is a pivot, and that its distribution function under the distribution with parameter  $\theta$  is given by

$$P_\theta(\theta^2 X_{(1)} \leq t) = \begin{cases} 1 - e^{-n\sqrt{t}} & \text{for } t \geq 0 \\ 0 & \text{for } t < 0. \end{cases}$$

- (b) Using this pivot from part (a) and for  $0 < \alpha < 1$ , find  $0 < c < d$  such that  $[cX_{(1)}^{-1/2}, dX_{(1)}^{-1/2}]$  is a confidence interval for  $\theta$  with confidence level  $1 - \alpha$

4. Let  $X_1, \dots, X_n$  be independent random variables with density

$$p_\theta(x) = \theta x^{\theta-1}, \quad x \in (0, 1),$$

where  $\theta > 0$ .

- (a) Compute the method of moments estimator for  $\theta$  based on the first moment.

The maximum likelihood estimator for  $\theta$  is given by

$$\hat{\theta}_{\text{mle}} = -\frac{n}{\sum \log X_i}.$$

- (b) Use this to show that the Likelihood Ratio test statistic for testing  $H_0 : \theta = 3$  against alternative  $H_1 : \theta \neq 3$  is given by

$$\lambda_n(X_1, \dots, X_n) = \left( \frac{\hat{\theta}_{\text{mle}}}{3} \right)^n \left( \prod_{i=1}^n X_i \right)^{\hat{\theta}_{\text{mle}} - 3}$$

In the book, a theorem is stated on the asymptotic distribution of  $2 \log \lambda_n$  under the null hypothesis. You may assume the conditions of this theorem are satisfied in the current setting.

- (c) Use this asymptotic distribution of  $2 \log \lambda_n$  to derive an approximate critical region for  $\lambda_n$  at significance level  $\alpha = 0.05$ .

5. Consider the simple linear regression model,

$$Y_i = \alpha + \beta x_i + e_i, \quad 1 \leq i \leq n,$$

where  $e_1, \dots, e_n$  are independent random variables with a  $\mathcal{N}(0, \sigma^2)$  distribution. The  $x_i$ 's are nonrandom. It is given that the maximum likelihood estimators for  $\beta$  and  $\alpha$  are given by

$$\hat{\beta} = \frac{\sum_{i=1}^n (x_i - \bar{x}_n)(Y_i - \bar{Y}_n)}{\sum_{i=1}^n (x_i - \bar{x}_n)^2} \text{ and } \hat{\alpha} = \bar{Y}_n - \hat{\beta} \bar{x}_n$$

where, as usual,  $\bar{x}_n = \frac{1}{n} \sum_{i=1}^n x_i$  and  $\bar{Y}_n = \frac{1}{n} \sum_{i=1}^n Y_i$

- (a) Write the estimator for  $\beta$  as linear combination of the observed  $Y_i$ 's, so as

$$\hat{\beta} = \sum_{i=1}^n \lambda_i Y_i$$

and specify  $\lambda_i$  for  $1 \leq i \leq n$ .

- (b) Show that  $\hat{\alpha}$  and  $\hat{\beta}$  are unbiased estimators for  $\alpha$  and  $\beta$  respectively.  
(c) Show that the Mean Squared Error of the estimator  $\hat{\beta}$  is given by

$$\text{MSE}(\hat{\beta}) = \frac{\sigma^2}{(n-1)s_x^2}$$

where  $s_x^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x}_n)^2$ .