

Probability and Statistics

Exercise sheet 13

Exercise 13.1

- (a) Let X_1, \dots, X_n be i.i.d. $\sim \text{Bernoulli}(\theta_0)$, for some unknown $\theta_0 \in \Theta = (0, 1)$. Take $\hat{\theta}_n = X_1$ as an estimator of θ_0 . We know that $T = T(X_1, \dots, X_n) = \sum_{i=1}^n X_i$ is sufficient for this model.

Let $\tilde{\theta}_n = E[\hat{\theta}_n | T] = E[X_1 | \sum_{i=1}^n X_i]$. Show that $\tilde{\theta}_n = \bar{X}_n$ and compute

$$\text{eff}(\tilde{\theta}_n, \hat{\theta}_n) = \frac{\text{MSE}(\hat{\theta}_n)}{\text{MSE}(\tilde{\theta}_n)}.$$

- (b) Let X_1, \dots, X_n be i.i.d. $\sim U([0, \theta_0])$, for some $\theta_0 \in \Theta = (0, +\infty)$. Consider $\hat{\theta}_n = 2X_1$ as an estimator of θ_0 . Let $T = T(X_1, \dots, X_n) = \max_{1 \leq i \leq n} X_i$. We have shown before that T is sufficient for this model. Let $\tilde{\theta}_n = E[\hat{\theta}_n | T] = 2E[X_1 | \max_{1 \leq i \leq n} X_i]$.

Remark: The random vector $(X_1, \max_i X_i)^T$ does not admit a density with respect to the Lebesgue measure on \mathbb{R}^2 since $P(X_1 = \max X_i) \neq 0$. Therefore, we will instead compute $\tilde{\theta}_n$ explicitly in the following indirect way.

- You are given the following result: Let X_1, \dots, X_n be i.i.d random variables with joint density $\prod_{i=1}^n f(x_i)$ with respect to Lebesgue measure on $(\mathbb{R}^n, \mathcal{B}_{\mathbb{R}^n})$. If one orders X_1, \dots, X_n in increasing order, we obtain the new random variables $X_{(1)} < \dots < X_{(n)}$, the so-called order statistics. For example, $X_{(1)} = \min_{1 \leq i \leq n} X_i$ and $X_{(n)} = \max_{1 \leq i \leq n} X_i = T$. The result says that for any $1 \leq i < j \leq n$, the random vector $(X_{(i)}, X_{(j)})^T$ is absolutely continuous with respect to Lebesgue measure on $(\mathbb{R}^2, \mathcal{B}_{\mathbb{R}^2})$ with density

$$g_{i,j}(s, t) = \frac{n!}{(i-1)!(j-1-i)!(n-j)!} f(s)f(t)[F(s)]^{i-1}[F(t)-F(s)]^{j-1-i}(1-F(t))^{n-j} \mathbb{1}_{s < t}$$

where F is the cdf corresponding to f .

If $j = n$, $i \in \{1, \dots, n-1\}$ and the X_i are once again i.i.d. $\sim U([0, \theta_0])$, compute explicitly the joint density $g_{i,n}$ for $(X_{(i)}, X_{(n)})$.

- Find the conditional density of $X_{(i)}$ given $X_{(n)}$, and use it to compute the conditional expectation $E[X_{(i)} | X_{(n)}]$.
Hint: Can you relate this conditional density to a well-known distribution, for which we know the expectation?
- Use a symmetry argument, and the fact that $\sum_{i=1}^n X_i = \sum_{i=1}^n X_{(i)}$, to show that $\tilde{\theta}_n = 2E[X_1 | X_{(n)}] = \frac{n+1}{n} X_{(n)}$.
- To conclude, show that $\hat{\theta}_n$ and $\tilde{\theta}_n$ are both unbiased and compute $\text{eff}(\tilde{\theta}_n, \hat{\theta}_n)$.

Exercise 13.2 Consider X_1, \dots, X_n i.i.d. $\sim \text{Exp}(\lambda)$, $\lambda \in \Theta = (0, +\infty)$. Recall that the pdf of $X_i \sim \text{Exp}(\lambda)$ is given by $f(x | \lambda) = \lambda e^{-\lambda x} \mathbb{1}_{x \in (0, +\infty)}$. We want to test $H_0 : \lambda = 1$ versus $H_1 : \lambda = 2$.

- (a) Apply the Neyman-Pearson Lemma to find a uniformly most powerful test of level α , based on $X = (X_1, \dots, X_n)^T$.

Hint: We recall that if Y_1, \dots, Y_n are $\overset{\text{iid}}{\sim} \text{Exp}(\lambda_0)$, then $\sum_{i=1}^n Y_i \sim G(n, \lambda_0)$.

(b) What is the power of the Neyman-Pearson test you found?

Hint: You can express your answer in terms of F_n and F_n^{-1} , the cdf and inverse cdf of a $G(n, 1)$ distribution.

(c) For $n = 10$, we observe the following sample:

1.009	0.132	0.384	0.360	0.206	0.588	0.872	0.398	0.339	1.079
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What decision do you take, if you want the level of the test to be equal to $\alpha = 0.05$? What about $\alpha = 0.01$?

Hint: The quantiles of the $G(10, 1)$ distribution of order 5% and 1% are 5.425 and 4.130, respectively.

Exercise 13.3 Again in the setup of exercise 2, it turns out that the Neyman-Pearson test you found in (a) is actually UMP of level α for testing $H_0 : \lambda = 1$ versus $H_1' : \lambda > 1$. More concretely, the same NP test is the most powerful among all tests of level α , for any $\lambda \in \Theta_1' = (1, +\infty)$, and not only for $\lambda \in \Theta_1 = \{2\}$.

Do you see why this is true?