

(1) Consider the distribution that has a p.d.f.

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

a. Compute the characteristic function for this distribution.

Also, compute the mean μ and variance σ^2 .

b. Now let X_1, X_2, \dots, X_n be i.i.d. from this distribution, and use a characteristic function argument in order to prove that

$$\sqrt{n} (\bar{X}_n - \mu) / \sigma \xrightarrow{L} Z \sim N(0,1).$$

c. Give a precise statements (no proof required!) of the Helly-Bray and continuity theorems. Then write an account of how these results are used to justify the technique that was used in part b. (above).

(2) Let X_1, X_2, \dots, X_n be i.i.d. from a distribution that has a finite 4th moment.

a. Derive the 2-dimensional limiting distribution of

$$\bar{X}_n = \frac{X_1 + X_2 + \dots + X_n}{n} \quad \text{and} \quad S_n^2 = \frac{\sum_{i=1}^n (X_i - \bar{X})^2}{n} .$$

b. Use the result in part a. to write out an (approximate) 95% confidence ellipsoid for the pair (μ, σ^2) .

c. The limiting distribution of S_n^2 , in particular, is *not* distribution free. Explain what this means along with its practical importance.

d. Finally, suppose the underlying distribution is Poisson, i.e.,

$$f_x(x; \lambda) = \begin{cases} \frac{\lambda^x e^{-\lambda}}{x!} & \lambda > 0, x = 0, 1, 2, \dots \\ 0 & \text{otherwise} \end{cases}$$

Thus the mean and variance are both equal to λ ; and, in this case, \bar{X}_n and S_n^2 are both unbiased (for λ). From the asymptotic point-of-view, which estimator is preferred and why?

Question I.

1. Define sufficiency and state the Neyman factorization criterion for it.

$(X_i, Y_i) \sim \text{iid } f$, $i=1, \dots, n$, where f is the uniform density on the triangle with vertices at $(0,0)$, $(0,\theta)$, $(\theta,0)$ ($\theta > 0$).

2. Are X_i and Y_i independent? uncorrelated?

3. Show that $T = \max_i (X_i + Y_i)$ is sufficient for θ .

4. Show that $X_i + Y_i$ has a triangular density (rising on $[0, 2\theta]$), and deduce the distribution of T .

5. What is the MLE $\hat{\theta}$ of θ ?

6. By what quick argument do you know $\hat{\theta}$ is biased? Is the bias removable?

7. Show that $\hat{\theta}$ has the same distribution as if it were the MLE on n iid data pairs on the square $[0, \theta] \times [0, \theta]$. Avoid messy calculation where possible.

8. What is the order of consistency of $\hat{\theta}$: \sqrt{n} ?; better?; worse? comment, skipping detailed calculation.

Question II.

1. Define admissibility and minimaxity of an estimation procedure for a statistical problem.

$X_i \sim \text{iid } \mathcal{N}(\theta, 1)$; $\theta \in \mathbb{R}$.

2. Is the sample mean \bar{X} admissible? minimax?
(brief comments). What about in higher dimensions?
3. If θ were known to be positive, would the sample mean be admissible? minimax?
4. Compute the Fisher information for θ .
5. Derive the Bayes estimator of θ , where $\theta \sim N(0, 1)$ a priori.
Is it admissible? minimax?
6. Perform an efficiency comparison of the Bayes estimator against \bar{X} .

Question III.

Let $X_1, X_2, \dots, X_n, \dots$ be i.i.d. from a $N(\theta, \theta)$ distribution in which $\theta > 0$ is unknown.

1. In general terms, describe an application in which this particular model might be of interest.
2. Compute the maximum likelihood estimator (MLE) of θ .
3. Obtain the asymptotic distribution of the MLE. Then write down a (large sample) 95% confidence interval for θ .
4. Two naive estimators of θ are given by

$$\hat{\theta}_1 = \bar{X} \quad \text{and} \quad \hat{\theta}_2 = \frac{\sum_{i=1}^n (X_i - \bar{X})^2}{n}.$$

Compare each of these estimators with the MLE by using the techniques of large sample theory.

5. Is \bar{X} complete? sufficient? Find a statistic that is both complete and sufficient.