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4. Maximum Likelihood Estimator for Curved Gaussian

(a)

1.0/1 point (graded)

Note: To avoid too much double jeopardy, the solution to part (a) will be available once you have either answered it correctly or reached the maximum number of attempts.

Let X_1, \dots, X_n be n i.i.d. random variables with distribution $\mathcal{N}(\theta, \theta)$ for some unknown $\theta > 0$.

Compute the maximum likelihood estimator $\hat{\theta}$ for θ in terms of the sample averages of the linear and quadratic means, i.e. \bar{X}_n and $\overline{X_n^2}$.

(Enter **barX_n** for \bar{X}_n and **bar(X_n^2)** for $\overline{X_n^2}$. Note that **barX_n^2** represents $(\bar{X}_n)^2$, and is **not** equal to **bar(X_n^2)** with the brackets.)

$$\hat{\theta} = (-1 + \sqrt{1 + 4 \cdot \text{bar}(X_n^2)}) \quad \checkmark$$

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STANDARD NOTATION

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You have used 3 of 3 attempts

(b)

4/4 points (graded)

We want to compute the asymptotic variance of $\hat{\theta}$ via two methods.

In this problem, we apply the Central Limit Theorem and the 1-dimensional Delta Method. We will compare this with the approach using the Fisher information next week.

First, compute the limit and asymptotic variance of $\overline{X_n^2}$.

The limit to which $\overline{X_n^2}$ converges in probability, also known as its **P-limit**, is

$$\overline{X_n^2} \xrightarrow[n \rightarrow \infty]{\mathbf{P}} \boxed{\theta^2 + \theta} \quad \checkmark$$

The asymptotic variance $V(\overline{X_n^2})$ of $\overline{X_n^2}$, which is equal to $\text{Var}(X_1^2)$, is

$$V(\overline{X_n^2}) = \text{Var}(X_1^2) = \boxed{4\theta^3 + 2\theta^2} \quad \checkmark$$

Now, write $\hat{\theta}$ as the function of $\overline{X_n^2}$ you found in part (a),

$$\hat{\theta} = g(\overline{X_n^2})$$

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and give its first derivative, $g'(x)$,

$$g'(x) = \frac{1}{\sqrt{1+4x}}$$

What can you conclude about the asymptotic variance $V(\hat{\theta})$ of $\hat{\theta}$?

$$V(\hat{\theta}) = \frac{2\theta^2}{1+2\theta}$$

STANDARD NOTATION

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You have used 2 of 3 attempts

✓ Correct (4/4 points)

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Part (b) asymptotic variance

discussion posted 22 minutes ago by [sandipan dey](#).

I really liked the problem very much and also interested to know about the thought-process to come up with a function $g(\cdot)$ s.t. $g(\theta^2 + \theta) = \theta$ structuring the problem.



It's nice to notice that the quadratic $\theta^2 + \theta = 1$ has the +ve root $\frac{-1+\sqrt{5}}{2} = g(1)$.

Question: Why asymptotic variance of \bar{X}_n^2 i.e. $Var(\bar{X}_n^2) = Var(X_1^2)$, whereas $Var(\bar{X}_n^2) = Var\left(\frac{1}{n} \cdot \sum_{i=1}^n X_i^2\right) = \frac{1}{n} Var(X_i^2)$? Is it because when $n \rightarrow \infty$ the (sampling) distribution of \bar{X}_n^2 tends to be same as the (population) distribution of X_i^2 ?

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