

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. \overline{X}_n and $\overline{X_n^2}$.

(Enter **barX_n** for \overline{X}_n and **bar(X_n^2)** for $\overline{X_n^2}$.)

$\hat{\theta} =$ ☐ Answer: (sqrt(4 * bar(X_n^2) + 1) - 1)/2

[STANDARD NOTATION](#)

Solution:

To compute the maximum likelihood estimator, we write the log likelihood and maximize it by setting its derivative to zero. First,

$$\begin{aligned}\ell_n(\theta) &= \sum_{i=1}^n \log \left[\frac{1}{\sqrt{2\pi\theta}} \exp \left(-\frac{(X_i - \theta)^2}{2\theta} \right) \right] \\ &= -\frac{n}{2} (\log(2) + \log(\pi) + \log(\theta)) - \sum_{i=1}^n \frac{(X_i - \theta)^2}{2\theta} \\ &= -\frac{n}{2} (\log(2) + \log(\pi) + \log(\theta)) - \sum_{i=1}^n \left[\frac{1}{2\theta} X_i^2 - X_i + \frac{1}{2} \theta \right].\end{aligned}$$

Differentiating yields

$$\frac{d}{d\theta} \ell(\theta) = -\frac{n}{2\theta} + \frac{1}{2\theta^2} \sum_{i=1}^n X_i^2 - \frac{n}{2},$$

which we set to zero to obtain the equation

$$\hat{\theta}^2 + \hat{\theta} - \frac{1}{n} \sum_{i=1}^n X_i^2 = 0.$$

Employing the quadratic formula and picking the result that gives a positive $\hat{\theta}$ then leads to

$$\hat{\theta} = -\frac{1}{2} + \frac{1}{2} \sqrt{4\overline{X_n^2} + 1}.$$

提交

你已经尝试了1次 (总共可以尝试3次)

(b)

2/4 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

$\overline{X_n^2} \xrightarrow[n \rightarrow \infty]{\mathbf{P}}$

theta^2 + theta

$\theta^2 + \theta$

☐ Answer: theta + theta^2

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

$V(\overline{X_n^2}) = \text{Var}(X_1^2) =$

2*theta^4+8*theta^3+4

$2 \cdot \theta^4 + 8 \cdot \theta^3 + 4 \cdot \theta^2$

☐ Answer: 2*theta^2*(2*theta + 1)

两个相减，我加起来了，太蠢了

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

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

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

$g'(x) =$

1/sqrt(4*x+1)

$\frac{1}{\sqrt{4 \cdot x + 1}}$

☐ Answer: 1/sqrt(4*x+1)

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

$V(\hat{\theta}) =$

1/(4*(theta^2 + theta)+1)

$\frac{1}{4 \cdot (\theta^2 + \theta) + 1} \cdot (2 \cdot \theta^4 + 8 \cdot \theta^3 + 4 \cdot \theta^2)$

☐ Answer: 2*theta^2/(2*theta + 1)

STANDARD NOTATION

Integrate[$\frac{x^4}{\sqrt{2 \pi \theta}} e^{-(x-\theta)^2/(2 \theta)}$, {x, -∞, ∞}, Assumptions → θ ≥ 0]

Solution:

First, by the Law of Large Numbers,

$$\overline{X_n^2} \xrightarrow[n \rightarrow \infty]{\mathbf{P}} \mathbb{E}[X_1^2] = \text{Var}(X_1) + \mathbb{E}[X_1]^2 = \theta + \theta^2.$$

Its asymptotic variance can be found by the Central Limit Theorem that gives us

$$\sqrt{n}(\overline{X_n^2} - (\theta + \theta^2)) \xrightarrow[n \rightarrow \infty]{(D)} \mathcal{N}(0, \text{Var}(X_1^2)),$$

and

$$\begin{aligned} \text{Var}(X_1^2) &= \mathbb{E}[X_1^4] - \mathbb{E}[X_1^2]^2 \\ &= \mathbb{E}[(\theta + \sqrt{\theta}Z)^4] - (\theta + \theta^2)^2 \end{aligned}$$

$$\begin{aligned}
 &= \theta^4 + 4\theta^3 \underbrace{\sqrt{\theta}\mathbb{E}[Z]}_{=0} + 6\theta^2 \underbrace{\theta\mathbb{E}[Z^2]}_{=0} + 4\theta \underbrace{\sqrt{\theta}^3\mathbb{E}[Z^3]}_{=0} + \underbrace{\theta^2\mathbb{E}[Z^4]}_{=3} - \theta^4 - 2\theta^3 + \theta^2 \\
 &= 2\theta^2(2\theta + 1),
 \end{aligned}$$

where $Z \sim \mathcal{N}(0, 1)$ is a standard Normal variable.

From the previous part, we get

$$g(x) = \frac{1}{2}(\sqrt{4x+1}-1),$$

so

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

Finally, by the Delta Method,

$$\sqrt{n}(g(\overline{X_n^2}) - g(\theta + \theta^2)) \overset{(D)}{\underset{n \rightarrow \infty}{\longrightarrow}} \mathcal{N}(0, 2\theta^2(2\theta + 1)g'(\theta + \theta^2)^2) = \mathcal{N}\left(0, \frac{2\theta^2}{2\theta + 1}\right).$$

提交

你已经尝试了3次（总共可以尝试3次）

☐ Answers are displayed within the problem

讨论

显示讨论

主题： Unit 3 Methods of Estimation:Homework 5: Maximum Likelihood Estimation / 4. Maximum Likelihood Estimator for Curved Gaussian