

Midterm Exam I

Estimation and Detection (Fall 2016)

* There are 4 problems in 2 pages.

* Total points: 105

1. (60 points) Let

$$x[n] = A + \omega[n], \quad n = 0, 1, \dots, N-1,$$

where $\omega[n] \sim \mathcal{N}(0, \sigma^2)$

(a) (5 points) Assume σ^2 is known. Obtain the CRLB and MVUE for A .

Sol:

see p.11-p.12 in Ch3.

$\frac{\sigma^2}{N}$ is CRLB and \bar{x} is the MVUE.

(b) (6 points) Followed (a), obtain the PDF for the MVUE.

Sol:

$$\hat{A} \sim \mathcal{N}\left(A, \frac{\sigma^2}{N}\right)$$

(c) (6 points) Followed (a), what is the CRLB for A^2 ?

Sol:

see p.21 in Ch3.

$$\frac{4A^2\sigma^2}{N}$$

(d) (15 points) Followed (a), let the MVUE be $g(x)$. Obtain the mean and variance for $(g(x))^2$. (Hint: If $x \sim \mathcal{N}(\mu, \sigma^2)$, $\mathbb{E}\{x^2\} = \mu^2 + \sigma^2$
 $\mathbb{E}\{x^4\} = \mu^4 + 6\mu^2\sigma^2 + 3\sigma^4$)

Sol:

see p.36 in Ch5.

$$\begin{aligned}\mathbb{E}\{\bar{x}^2\} &= A^2 + \frac{\sigma^2}{N} \\ \text{var}\{\bar{x}^2\} &= \mathbb{E}\{(\bar{x}^2)^2\} - (\mathbb{E}\{\bar{x}^2\})^2 \\ &= A^4 + 6A^2\frac{\sigma^2}{N} + 3\left(\frac{\sigma^2}{N}\right)^2 - \left(A^4 + 2A^2\frac{\sigma^2}{N} + \frac{\sigma^4}{N^2}\right) \\ &= 4A^2\frac{\sigma^2}{N} + 2\frac{\sigma^4}{N^2} \\ &= \frac{\sigma^2}{N}\left(4A^2 + 2\frac{\sigma^2}{N}\right)\end{aligned}$$

(e) (5 points) Followed (d), is $(g(x))^2$ the MVUE for A^2 ?

Sol:

No.

$$\mathbb{E}\{\bar{x}^2\} = A^2 + \frac{\sigma^2}{N} \neq A^2$$

is biased.

(f) (8 points) Followed (d), in what condition the estimator $(g(x))^2$ can asymptotically achieve the CRLB and tend to be the MVUE?

Sol:

When $N \rightarrow \infty$ or $\sigma^2 \rightarrow 0$

$$\mathbb{E}\{\bar{x}^2\} = A^2 + \frac{\sigma^2}{N} \rightarrow A^2 \text{ asymptotically unbiased}$$

$$\text{var}\{\bar{x}^2\} = \frac{\sigma^2}{N}(4A^2 + \frac{\sigma^2}{N}) \rightarrow \frac{4A^2\sigma^2}{N} \text{ asymptotically achieve CRLB}$$

- (g) (15 points) Assume now both A and σ^2 are unknown. Obtain the sufficient statistics for (A, σ^2) . Assuming completeness, using these sufficient statistics to obtain the MVUE.

Sol:

see p.34-p.36 in Ch5.

2. (15 points) Given the following system model:

$$x[n] = A + Bn + Cn^2 + \omega[n], \quad n = 0, 1, 2$$

Answer the following questions with “explanations” or proof.

- (a) (8 points) Obtain the MVUEs for A,B and C.

Sol:

Let

$$\theta = \begin{bmatrix} A \\ B \\ C \end{bmatrix}, \mathbf{H} = \begin{bmatrix} 1 & 0 & 0 \\ 1 & 1 & 1 \\ 1 & 2 & 4 \end{bmatrix}$$

and

$$x = \mathbf{H}\theta + \omega$$

$$\hat{\theta} = \begin{bmatrix} \hat{A} \\ \hat{B} \\ \hat{C} \end{bmatrix} = \mathbf{H}^{-1}x = \begin{bmatrix} 1 & 0 & 0 \\ -1.5 & 2 & -0.5 \\ 0.5 & -1 & 0.5 \end{bmatrix} x$$

(b) (7 points) Indicate the minimum variance of these estimators.

Sol:

$$\begin{aligned}
 \text{minimum variance} &= \mathcal{I}^{-1}(\theta) = \sigma^2(\mathbf{H}^T \mathbf{H})^{-1} \\
 \text{var}\{\hat{A}\} &\geq \sigma^2 \\
 \text{var}\{\hat{B}\} &\geq 6.5\sigma^2 \\
 \text{var}\{\hat{C}\} &\geq 1.5\sigma^2
 \end{aligned} \tag{1}$$

3. (15 points) Given i.i.d observations $x[n]$ for $n = 0, 1, \dots, N - 1$, where $x[n]$ has the following PDF

$$p(x; \sigma^2) = \begin{cases} \frac{x}{\sigma^2} \exp \left\{ -\frac{1}{2} \frac{x^2}{\sigma^2} \right\} & , x > 0 \\ 0 & , x < 0 \end{cases} ,$$

(a) (10 points) Find a sufficient statistic for σ^2 .

Sol:

$$\begin{aligned}
 p(\mathbf{x}; \sigma^2) &= \prod_{n=0}^{N-1} u[x[n]] \frac{x[n]}{\sigma^2} \exp \left[\frac{-x^2[n]}{2\sigma^2} \right] \\
 &= \frac{1}{\sigma^{2N}} \exp \left[-\frac{1}{2\sigma^2} \sum_{n=0}^{N-1} x^2[n] \right] u[\min(x[n])] \prod_{n=0}^{N-1} x[n]
 \end{aligned}$$

From Neyman-fisher factorization ,the sufficient statistic for σ^2 is

$$\mathbf{T}(\mathbf{x}) = \sum_{n=0}^{N-1} x^2[n]$$

- (b) (5 points) Find a MVU estimator for σ^2 . (assuming the sufficient statistic found in (a) is complete)

Sol:

Find some function g so that $\hat{\sigma}^2 = g(\mathbf{T})$ is an unbiased estimator of σ^2 .

$$\begin{aligned} \because E[\mathbf{T}(\mathbf{x})] &= \sum_{n=0}^{N-1} E\{x^2[n]\} = 2N\sigma^2 \\ \therefore \text{Let } g(\mathbf{T}(\mathbf{x})) &= \frac{1}{2N}\mathbf{T}(\mathbf{x}) \quad \text{such that} \quad E\{g(\mathbf{T}(\mathbf{x}))\} = \sigma^2 \end{aligned}$$

4. (15 points) Given the following system model

$$x[n] = A \cos(2\pi f_0 n + \phi) + \omega[n], \quad n = 0, 1, \dots, N-1,$$

where $\omega[n]$ is WGN with zero mean and variance σ^2 . A and f_0 is assumed known.

- (a) (7 points) Find the CRLB for $\text{var}\{\phi\}$.

Sol: The PDF is

$$\begin{aligned} p(\mathbf{x}; \phi) &= \prod_{n=0}^{N-1} \frac{1}{\sqrt{2\pi\sigma^2}} \exp \left[\frac{-(x[n] - A \cos(2\pi f_0 n + \phi))^2}{2\sigma^2} \right] \\ &= \frac{1}{(2\pi\sigma^2)^{\frac{N}{2}}} \exp \left[-\frac{1}{2\sigma^2} \sum_{n=0}^{N-1} (x[n] - A \cos(2\pi f_0 n + \phi))^2 \right] \end{aligned}$$

The log-likelihood function

$$\ln p(\mathbf{x}; \phi) = -\frac{N}{2} \ln(2\pi\sigma^2) + \left[-\frac{1}{2\sigma^2} \sum_{n=0}^{N-1} (x[n] - A \cos(2\pi f_0 n + \phi))^2 \right]$$

Differentiation the log-likelihood function

$$\frac{\partial \ln p(\mathbf{x}; \phi)}{\partial \phi} = -\frac{A}{\sigma^2} \sum_{n=0}^{N-1} \left[x[n] \sin(2\pi f_0 n + \phi) - \frac{A}{2} \sin(4\pi f_0 n + 2\phi) \right]$$

and

$$\frac{\partial^2 \ln p(\mathbf{x}; \phi)}{\partial^2 \phi} = -\frac{A}{\sigma^2} \sum_{n=0}^{N-1} [x[n] \cos(2\pi f_0 n + \phi) - A \cos(4\pi f_0 n + 2\phi)]$$

From Fisher information and CRLB, we know

$$\begin{aligned} I(\phi) &= -\mathbb{E} \left[\frac{\partial^2 \ln p(\mathbf{x}; \phi)}{\partial^2 \phi} \right] \\ &= \frac{A^2}{\sigma^2} \sum_{n=0}^{N-1} \left[\frac{1}{2} - \frac{1}{2} \cos(4\pi f_0 n + 2\phi) \right] \\ &\approx \frac{NA^2}{2\sigma^2} \\ \text{var}(\hat{\phi}) &\geq \frac{1}{I(\phi)} \\ &\geq \frac{2\sigma^2}{NA^2} \end{aligned}$$

- (b) (8 points) Now letting $\phi = 0$, assume f_0 and σ^2 are known. Find the MVU estimator of A . (Hint: find sufficient statistic and assumed it is complete)

Sol:

$$\begin{aligned}
 p(\mathbf{x}; \phi) &= \frac{1}{(2\pi\sigma^2)^{\frac{N}{2}}} \exp \left[-\frac{1}{2\sigma^2} \sum_{n=0}^{N-1} (x[n] - A \cos(2\pi f_0 n))^2 \right] \\
 &= \frac{1}{(2\pi\sigma^2)^{\frac{N}{2}}} \exp \left[-\frac{1}{2\sigma^2} \left(\sum_{n=0}^{N-1} x[n]^2 - 2 \sum_{n=0}^{N-1} x[n] A \cos(2\pi f_0 n) \right. \right. \\
 &\quad \left. \left. + \sum_{n=0}^{N-1} A^2 \cos^2(2\pi f_0 n) \right) \right]
 \end{aligned}$$

Using Neyman-fisher factorization ,the sufficient statistic is

$$\mathbf{T}(\mathbf{x}) = \sum_{n=0}^{N-1} x[n] \cos(2\pi f_0 n)$$

Find some function g so that $\hat{A} = g(\mathbf{T})$ is an unbiased estimator of A

$$\begin{aligned}
 \because E[\mathbf{T}(\mathbf{x})] &= \sum_{n=0}^{N-1} A \cos^2(2\pi f_0 n) \\
 \therefore \text{Let } g(\mathbf{T}(\mathbf{x})) &= \frac{\sum_{n=0}^{N-1} x[n] \cos(2\pi f_0 n)}{\sum_{n=0}^{N-1} \cos^2(2\pi f_0 n)} \quad \text{such that } E\{g(\mathbf{T}(\mathbf{x}))\} = A
 \end{aligned}$$