

1. This problem illustrates what happens to an unbiased estimator when it undergoes a non-linear transformation. Consider the observations

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

where, $w[n]$ is the White Gaussian Noise (WGN). The parameter A can take any value in the interval $-\infty < A < \infty$. If we chose to estimate the unknown parameter $\theta = A^2$ by,

$$\hat{\theta} = \left(\frac{1}{N} \sum_{n=0}^{N-1} x[n] \right)^2, \quad (2)$$

can we say that the estimator is unbiased? What happens as $N \rightarrow \infty$?

2. Consider the observations

$$x[n] = A + w[n], \quad n = 0, 1, \dots, N-1 \quad (3)$$

where, $w[n]$ is the White Gaussian Noise (WGN). Lets assume that in addition to A , the value of σ^2 is also unknown. We wish to estimate the vector parameter

$$\hat{\boldsymbol{\theta}} = \begin{bmatrix} A \\ \sigma^2 \end{bmatrix}. \quad (4)$$

Is the estimator

$$\hat{\boldsymbol{\theta}} = \begin{bmatrix} \hat{A} \\ \hat{\sigma}^2 \end{bmatrix} = \begin{bmatrix} \frac{1}{N} \sum_{n=0}^{N-1} x[n] \\ \frac{1}{N-1} \sum_{n=0}^{N-1} (x[n] - \hat{A})^2 \end{bmatrix} \quad (5)$$

unbiased?

3. We observe two samples of a DC level in *correlated* Gaussian noise

$$x[0] = A + w[0] \quad (6)$$

$$x[1] = A + w[1] \quad (7)$$

where $\mathbf{w} = [w[0] \ w[1]]^T$ is zero mean with covariance matrix

$$\mathbf{C} = \sigma^2 \begin{bmatrix} 1 & \rho \\ \rho & 1 \end{bmatrix}. \quad (8)$$

The parameter ρ is the correlation coefficient between $w[0]$ and $w[1]$. Compute the CRLB for A and compare it to the case when $w[n]$ is WGN or $\rho = 0$. Also, explain what happens when $\rho \rightarrow \pm 1$. Finally, comment on the additivity property of the Fisher information for non-independent observations.

4. Consider a generalization of the line fitting problem as described in Example 3.7, termed *polynomial or curve fitting*. The data model is

$$x[n] = \sum_{k=0}^{p-1} A_k n^k + w[n] \quad (9)$$

for $n = 0, 1, \dots, N-1$. As before, $w[n]$ is WGN with variance σ^2 . It is desired to estimate $\{A_0, A_1, \dots, A_{p-1}\}$. Find the Fisher information matrix for this problem.

5. Independent bivariate Gaussian samples $\{\mathbf{x}[0], \mathbf{x}[1], \dots, \mathbf{x}[N-1]\}$ are observed. Each observation is a 2×1 vector which is distributed as $\mathbf{x}[n] \sim \mathcal{N}(\mathbf{0}, \mathbf{C})$ and

$$\mathbf{C} = \begin{bmatrix} 1 & \rho \\ \rho & 1 \end{bmatrix}. \quad (10)$$

Find the CRLB for the correlation coefficient ρ .

6. A line array of antennas with $d = \lambda/2$ spacing for a 1 MHz electromagnetic signal is to be designed to yield a 5° standard deviation at $\beta = 90^\circ$. If the SNR η is 0 dB, comment on the feasibility of the requirement. Assume that the line array is to be mounted on the wing of an aircraft. Use $c = 3 \times 10^8$ m/s.

7. Consider the observation matrix

$$\mathbf{H} = \begin{bmatrix} 1 & 1 \\ 1 & 1 \\ 1 & 1 + \epsilon \end{bmatrix} \quad (11)$$

where ϵ is small. Compute $(\mathbf{H}^T \mathbf{H})^{-1}$ and examine what happens as $\epsilon \rightarrow 0$. If $\mathbf{x} = [2 \ 2 \ 2]^T$, find the MVU estimator and describe what happens as $\epsilon \rightarrow 0$.

8. In the linear model it is desired to estimate the signal $\mathbf{s} = \mathbf{H}\boldsymbol{\theta}$. If an MVU estimator of $\boldsymbol{\theta}$ is found, then the signal may be estimated as $\hat{\mathbf{s}} = \mathbf{H}\hat{\boldsymbol{\theta}}$. What is the PDF of $\hat{\mathbf{s}}$?
9. In practice we sometimes encounter the "linear model" $\mathbf{x} = \mathbf{H}\boldsymbol{\theta} + \mathbf{w}$ but *with \mathbf{H} composed of random variables*. Suppose we ignore this difference and use our usual estimator

$$\hat{\boldsymbol{\theta}} = (\mathbf{H}^T \mathbf{H})^{-1} \mathbf{H}^T \mathbf{x}, \quad (12)$$

where we assume that the particular realization of \mathbf{H} is known to us. Show that if \mathbf{H} and \mathbf{w} are independent, the mean and covariance of $\hat{\boldsymbol{\theta}}$ are

$$E(\hat{\boldsymbol{\theta}}) = \boldsymbol{\theta} \quad (13)$$

$$\mathbf{C}_{\hat{\boldsymbol{\theta}}} = \sigma^2 E_H [(\mathbf{H}^T \mathbf{H})^{-1}] \quad (14)$$

where E_H denotes the expectation with respect to the PDF of \mathbf{H} . What happens if the independence assumption is not made?

10. If $x[n] = A + w[n]$ for $n = 0, 1, 2, \dots, N-1$ are observed and

$$w = [w[0] \ w[1] \ w[2] \ \dots \ w[N-1]]^T \sim N(0, C) \quad (15)$$

find the CRLB for A. Does an efficient estimator exist and if so, what is its variance?

11. For a 2×2 Fisher information matrix $I = \begin{bmatrix} a & b \\ b & c \end{bmatrix}$ which is positive definite, show that

$$[I^{-1}(\theta)]_{11} = \frac{c}{ac - b^2} \geq \frac{1}{a} = \frac{1}{[I(\theta)]_{11}} \quad (16)$$

What does this say about estimating a parameter when a second parameter is either known or unknown? When does equality hold and why?