

Technique for Numerical Computation of Cramér-Rao Bound using MATLAB

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The purpose of this note is to provide an effective means to compute the **Cramér-Rao lower bound (CRLB)** numerically with the use of symbolic computation in MATLAB.

CRLB is the performance bound in terms of **minimum achievable variance** provided by any **unbiased** estimators.

For deterministic parameter estimation, its derivation requires knowledge of the noise **probability density function (PDF)** which must have **closed-form** so that **analytical differentiation** is allowed.

Although there are other variance bounds, CRLB is **simplest**.

Suppose the PDF of $\mathbf{r} = \mathbf{f}(\mathbf{x}) + \mathbf{w}$ is $p(\mathbf{r}; \mathbf{x})$. Here, \mathbf{f} is a known linear/nonlinear function, $\mathbf{x} = [x_1 \ x_2 \ \cdots \ x_L]^T \in \mathbb{R}^L$ is the parameter vector, and \mathbf{w} is noise vector.

The CRLB for \mathbf{x} can be obtained in two steps:

- Compute the **Fisher information matrix** $\mathbf{I}(\mathbf{x})$ where

$$\mathbf{I}(\mathbf{x}) = \begin{bmatrix} -E \left\{ \frac{\partial^2 \ln p(\mathbf{r}; \mathbf{x})}{\partial^2 x_1} \right\} & -E \left\{ \frac{\partial^2 \ln p(\mathbf{r}; \mathbf{x})}{\partial x_1 \partial x_2} \right\} & \cdots & -E \left\{ \frac{\partial^2 \ln p(\mathbf{r}; \mathbf{x})}{\partial x_1 \partial x_L} \right\} \\ -E \left\{ \frac{\partial^2 \ln p(\mathbf{r}; \mathbf{x})}{\partial x_2 \partial x_1} \right\} & -E \left\{ \frac{\partial^2 \ln p(\mathbf{r}; \mathbf{x})}{\partial^2 x_2} \right\} & \cdots & -E \left\{ \frac{\partial^2 \ln p(\mathbf{r}; \mathbf{x})}{\partial x_2 \partial x_L} \right\} \\ \vdots & \vdots & \vdots & \vdots \\ -E \left\{ \frac{\partial^2 \ln p(\mathbf{r}; \mathbf{x})}{\partial x_L \partial x_1} \right\} & \cdots & \cdots & -E \left\{ \frac{\partial^2 \ln p(\mathbf{r}; \mathbf{x})}{\partial^2 x_L} \right\} \end{bmatrix}$$

- CRLB for x_l is the (l, l) entry of $\mathbf{I}^{-1}(\mathbf{x})$, $l = 1, 2, \dots, L$.

Consider $r[n] = A + w[n]$, $n = 1, \dots, N$, where $A \in \mathbb{R}$ is the parameter to be estimated and $w[n] \in \mathbb{R}$ is zero-mean white Gaussian noise with variance σ_w^2 . The CRLB is derived as:

$$\begin{aligned}
 p(\mathbf{r}; A) &= \frac{1}{(2\pi\sigma_w^2)^{N/2}} e^{-\frac{1}{2\sigma_w^2} \sum_{n=1}^N (r[n]-A)^2} \\
 \Rightarrow \ln p(\mathbf{r}; A) &= -\ln((2\pi\sigma_w^2)^{N/2}) - \frac{1}{2\sigma_w^2} \sum_{n=1}^N (r[n] - A)^2 \\
 \Rightarrow \frac{\partial^2 \ln p(\mathbf{r}; A)}{\partial^2 A} &= -\frac{N}{\sigma_w^2} \\
 \Rightarrow \mathbf{I}(A) &= -E \left\{ \frac{\partial^2 \ln p(\mathbf{r}; A)}{\partial^2 A} \right\} = \frac{N}{\sigma_w^2} \\
 \Rightarrow \mathbf{I}^{-1}(A) &= \frac{\sigma_w^2}{N}
 \end{aligned}$$

The CRLB can be obtained easily in closed-form for this simple example.

How about for more complicated problems? We use another simple example for its numerical calculation as follows.

Consider

$$r[n] = A_1 e^{j(\omega_1 n + \phi_1)} + A_2 e^{j(\omega_2 n + \phi_2)} + w[n], \quad n = 1, \dots, N$$

where $\mathbf{x} = [\omega_1 \ \omega_2 \ \phi_1 \ \phi_2 \ A_1 \ A_2]^T \in \mathbb{R}^6$ and $w[n] \in \mathbb{C}$ is zero-mean white complex Gaussian noise with variance σ_w^2 .

Ignoring the irrelevant term, the log PDF is:

$$\begin{aligned} \ln p(\mathbf{r}; \mathbf{x}) &= -\frac{1}{\sigma_w^2} \sum_{n=1}^N \left(r[n] - A_1 e^{j(\omega_1 n + \phi_1)} - A_2 e^{j(\omega_2 n + \phi_2)} \right) \\ &\quad \left(r^*[n] - A_1 e^{-j(\omega_1 n + \phi_1)} - A_2 e^{-j(\omega_2 n + \phi_2)} \right) \end{aligned}$$

We can see that the main computation is to find the second-order derivatives for:

$$p = \left(r[n] - A_1 e^{j(\omega_1 n + \phi_1)} - A_2 e^{j(\omega_2 n + \phi_2)} \right) \left(r^*[n] - A_1 e^{-j(\omega_1 n + \phi_1)} - A_2 e^{-j(\omega_2 n + \phi_2)} \right)$$

With the use of symbolic computation of MATLAB, we define:

```
syms w1 w2 p1 p2 A1 A2 n s1 s2 p;
p = (s1 - A1 * exp(j * w1 * n) * exp(j * p1) -
A2 * exp(j * w2 * n) * exp(j * p2)) * (s2 - A1 * exp(-
j * w1 * n) * exp(-j * p1) - A2 * exp(-j * w2 * n) * exp(-j * p2));
```

Here, the first 6 symbols correspond to x , and the remaining symbols are time index (n), $r[n]$ ($s1$), $r^*[n]$ ($s2$) and p (p).

For example, in computing $\frac{\partial^2 p}{\partial^2 \omega_1}$ in analytical form, we use:

```
p11=simple(diff(diff(p,w1),w1))
```

Note that `simple` is used to simplify the resultant expression.

In a similar manner, we compute all other required second-order derivatives. Note also that we only need to determine the upper triangular elements of $\mathbf{I}(\mathbf{x})$ as the matrix is symmetric.

For detail, please refer to `comp_diff.m` or see below:

http://www.ee.cityu.edu.hk/~hcs0/comp_diff.m

comp_diff.m

%Parameter vector consists of symbols:

%frequencies: w1 and w2;

%initial phases: p1 and p2;

%real-valued amplitudes: A1 and A2

%Other defined symbols are:

%r[n] and its conjugate: s1 and s2;

%time index: n;

%log PDF: p

%pij stands for the (i,j) entry of the FIM (individual and without expectation)

%e.g., p11 corresponds to differentiation of p w.r.t. w1 twice

syms w1 w2 p1 p2 A1 A2 n s1 s2 p;

p=(s1-A1*exp(j*w1*n)*exp(j*p1)-A2*exp(j*w2*n)*exp(j*p2))*(s2-A1*exp(-j*w1*n)*exp(-j*p1)-A2*exp(-j*w2*n)*exp(-j*p2));

p11=simple(diff(diff(p,w1),w1))

p12=simple(diff(diff(p,w1),w2))

p13=simple(diff(diff(p,w1),p1))

p14=simple(diff(diff(p,w1),p2))

p15=simple(diff(diff(p,w1),A1))

p16=simple(diff(diff(p,w1),A2))

p22=simple(diff(diff(p,w2),w2))

p23=simple(diff(diff(p,w2),p1))

p24=simple(diff(diff(p,w2),p2))

p25=simple(diff(diff(p,w2),A1))

p26=simple(diff(diff(p,w2),A2))

p33=simple(diff(diff(p,p1),p1))

p34=simple(diff(diff(p,p1),p2))

p35=simple(diff(diff(p,p1),A1))

p36=simple(diff(diff(p,p1),A2))

p44=simple(diff(diff(p,p2),p2))

p45=simple(diff(diff(p,p2),A1))

p46=simple(diff(diff(p,p2),A2))

p55=simple(diff(diff(p,A1),A1))

p56=simple(diff(diff(p,A1),A2))

p66=simple(diff(diff(p,A2),A2))

Then we copy the resultant expressions to another MATLAB file for $\mathbf{I}(\mathbf{x})$ computation.

Here, we assign:

$$s(n) = A1 * \exp(j * w1 * n) * \exp(j * p1) + A2 * \exp(j * w2 * n) * \exp(j * p2);$$

and replace $s1$ and $s2$ by $s(n)$ and $\text{conj}(s(n))$, respectively. We need to assign \mathbf{x} for numerical computation, and without loss of generality, here all parameters are set to one.

Note also that the expected value of $r[n]$ is $s(n)$.

For detail, please refer to `comp_fim.m` or see below:

http://www.ee.cityu.edu.hk/~hcso/comp_fim.m

comp_fim.m

```
%Finding FIM for two cisoids
%As an illustration:
%A1=A2=w1=w2=p1=p2=1
%Observation length is N=100
%Standard derivation of zero-mean white noise power is sigma=1

N=100; % data length
sigma=1; % noise standard deviation
fim=zeros(6,6); %initialize FIM
A1=1;
A2=1;
w1=1;
w2=2;
p1=1;
p2=2;
for n=1:N

s(n)=A1*exp(j*w1*n)*exp(j*p1)+A2*exp(j*w2*n)*exp(j*p2);

fim(1,1)=fim(1,1) + A1*n^2*(-2*A2*cos(w1*n+p1-w2*n-p2)+exp(i*(w1*n+p1))*conj(s(n))+exp(-i*(w1*n+p1))*s(n));
fim(1,2)=fim(1,2) +2*A1*n^2*A2*cos(w1*n+p1-w2*n-p2);
fim(1,3)=fim(1,3) +A1*n*(-2*A2*cos(w1*n+p1-w2*n-p2)+exp(i*(w1*n+p1))*conj(s(n))+exp(-i*(w1*n+p1))*s(n));
fim(1,4)=fim(1,4) +2*A1*n*A2*cos(w1*n+p1-w2*n-p2);
fim(1,5)=fim(1,5)-n*(2*A2*sin(w1*n+p1-w2*n-p2)+i*conj(s(n))*exp(i*(w1*n+p1))-i*s(n)*exp(-i*(w1*n+p1)));
fim(1,6)=fim(1,6)-2*n*A1*sin(w1*n+p1-w2*n-p2);

fim(2,2)=fim(2,2) +A2*n^2*(-2*A1*cos(w1*n+p1-w2*n-p2)+exp(i*(w2*n+p2))*conj(s(n))+exp(-i*(w2*n+p2))*s(n));
fim(2,3)=fim(2,3)+2*A1*n*A2*cos(w1*n+p1-w2*n-p2);
fim(2,4)=fim(2,4)+A2*n*(-2*A1*cos(w1*n+p1-w2*n-p2)+exp(i*(w2*n+p2))*conj(s(n))+exp(-i*(w2*n+p2))*s(n));
fim(2,5)=fim(2,5)+2*A2*n*sin(w1*n+p1-w2*n-p2);
fim(2,6)=fim(2,6)-n*(-2*A1*sin(w1*n+p1-w2*n-p2)+i*conj(s(n))*exp(i*(w2*n+p2))-i*s(n)*exp(-i*(w2*n+p2)));

fim(3,3)=fim(3,3)+A1*(-2*A2*cos(w1*n+p1-w2*n-p2)+exp(i*(w1*n+p1))*conj(s(n))+exp(-i*(w1*n+p1))*s(n));
fim(3,4)=fim(3,4)+2*A1*A2*cos(w1*n+p1-w2*n-p2);
fim(3,5)=fim(3,5)-2*A2*sin(w1*n+p1-w2*n-p2)-i*conj(s(n))*exp(i*(w1*n+p1))+i*s(n)*exp(-i*(w1*n+p1));
fim(3,6)=fim(3,6)-2*A1*sin(w1*n+p1-w2*n-p2);
```

```

fim(4,4)=fim(4,4)+A2*(-2*A1*cos(w1*n+p1-w2*n-p2)+exp(i*(w2*n+p2))*conj(s(n))+exp(-i*(w2*n+p2))*s(n));
fim(4,5)=fim(4,5)+2*A2*sin(w1*n+p1-w2*n-p2);
fim(4,6)=fim(4,6)+2*A1*sin(w1*n+p1-w2*n-p2)-i*conj(s(n))*exp(i*(w2*n+p2))+i*s(n)*exp(-i*(w2*n+p2));

fim(5,5)=fim(5,5)+2;
fim(5,6)=fim(5,6)+2*cos(w1*n+p1-w2*n-p2);

fim(6,6)=fim(6,6)+2;

end

fim(2,1)=fim(1,2);

fim(3,1)=fim(1,3);
fim(3,2)=fim(2,3);

fim(4,1)=fim(1,4);
fim(4,2)=fim(2,4);
fim(4,3)=fim(3,4);

fim(5,1)=fim(1,5);
fim(5,2)=fim(2,5);
fim(5,3)=fim(3,5);
fim(5,4)=fim(4,5);

fim(6,1)=fim(1,6);
fim(6,2)=fim(2,6);
fim(6,3)=fim(3,6);
fim(6,4)=fim(4,6);
fim(6,5)=fim(5,6);

fim=fim./sigma^2;

```