

## EE5147 Modern Spectral Analysis

### Homework Assignment #2

#### Notice

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- Due at 9:00pm, April 12, 2022 (Tuesday) =  $T_d$  for the electronic copy of your solution.
- Please submit your solution to NTU COOL (<https://cool.ntu.edu.tw/courses/11382>)
- Please justify your answers. If the problem begins with “Show that” or “Prove”, we expect a **rigorous proof**.
- All the figures should include labels for the horizontal and vertical axes, a title for a short description, and grid lines. Add legends and different line styles if there are multiple curves in one plot.
- No extensions, unless granted by the instructor one day before  $T_d$ .
- [MIK2005]: D. G. Manolakis, V. K. Ingle, and S. M. Kogon, *Statistical and Adaptive Signal Processing Spectral Estimation, Signal Modeling, Adaptive Filtering, and Array Processing*, Artech House, 2005.
- [SM2005]: P. Stoica and R. Moses, *Spectral Analysis of Signals*, Upper Saddle River, N.J.: Pearson/Prentice Hall, 2005.

#### Problems

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1. (10 points) Consider the following random process

$$x(n) = A_1 e^{j2\pi f_1 n} + A_2 e^{j2\pi f_2 n} + w(n), \quad (1)$$

where  $A_1$ ,  $A_2$ , and  $w(n)$  are jointly complex circularly-symmetric Gaussian distributed with zero mean. Furthermore, we assume

$$\mathbb{E}[A_p A_q^*] = \begin{cases} \sigma^2, & \text{if } p = q, \\ \frac{1}{3}\sigma^2, & \text{if } p \neq q. \end{cases}, \quad \mathbb{E}[w(n)w^*(n-k)] = \sigma_w^2 \delta(k), \quad \mathbb{E}[A_p w^*(n)] = 0. \quad (2)$$

The frequencies  $f_1$  and  $f_2$  are deterministic. We assume that  $f_1 \neq f_2$ .

- (a) (3 points) Find the mean-value function of  $x(n)$ .
- (b) (5 points) Find the autocorrelation function of  $x(n)$ .
- (c) (2 points) Determine whether  $x(n)$  is wide-sense stationary.

2. (10 points) We consider the exponentially damped sinusoidal components plus noise of the following form

$$x(n) = \sum_{p=1}^P \alpha_p e^{(\xi_p + j2\pi f_p)n} + w(n). \quad (3)$$

In this model, the quantities  $P$ ,  $\alpha_p$ ,  $f_p$ , and  $w(n)$  follow the same assumptions on page 9 of 04\_Harmonic\_MVDR.pdf. Furthermore, the damping parameter  $\xi_p$  is deterministic and negative ( $\xi_p < 0$ ). Let

$$\mathbf{x}(n) = \begin{bmatrix} x(n) \\ x(n+1) \\ \vdots \\ x(n+M-1) \end{bmatrix}, \quad (4)$$

be the vector form of (3). We assume that  $P < M$ .

- (a) (5 points) Express  $\mathbf{x}(n)$  in the following form

$$\mathbf{x}(n) = \mathbf{V}\mathbf{s}(n) + \mathbf{w}(n), \quad (5)$$

where the matrix  $\mathbf{V}$  is deterministic and the entries of the first row of  $\mathbf{V}$  are 1. Specify  $\mathbf{V}$ ,  $\mathbf{s}(n)$ , and  $\mathbf{w}(n)$ .

- (b) (5 points) Illustrate the eigenvalues of  $\mathbf{R} \triangleq \mathbb{E}[\mathbf{x}(n)\mathbf{x}^H(n)]$ .

3. (10 points) Consider an AR process  $x(n)$  that is produced by filtering unit variance white circularly-symmetric Gaussian noise  $w(n)$  with the system equation.

$$x(n) + a_1 x(n-1) = w(n), \quad (6)$$

where  $|a_1| < 1$ . Find the MVDR spectrum of  $x(n)$  for  $M = 2$ .

4. (10 points) Suppose the MVDR spectrum of a WSS random process  $x(n)$  is given by

$$\hat{S}_{x,\text{MVDR}}(e^{j2\pi f}) = \frac{8}{3 - \cos(2\pi f)}. \quad (7)$$

We assume that  $M = 2$ . Find the autocorrelations  $r_x(0)$  and  $r_x(1)$  that produces  $\hat{S}_{x,\text{MVDR}}(e^{j2\pi f})$ .

5. (15 points) Let  $\mathbf{c} = [c_0, c_1, \dots, c_{M-1}]^T$  be the linear combination coefficients for the MVDR spectrum estimation. The output  $y(n) = \mathbf{c}^H \mathbf{x}(n)$ , where

$$\mathbf{x}(n) = \begin{bmatrix} x(n) \\ x(n+1) \\ \vdots \\ x(n+M-1) \end{bmatrix}. \quad (8)$$

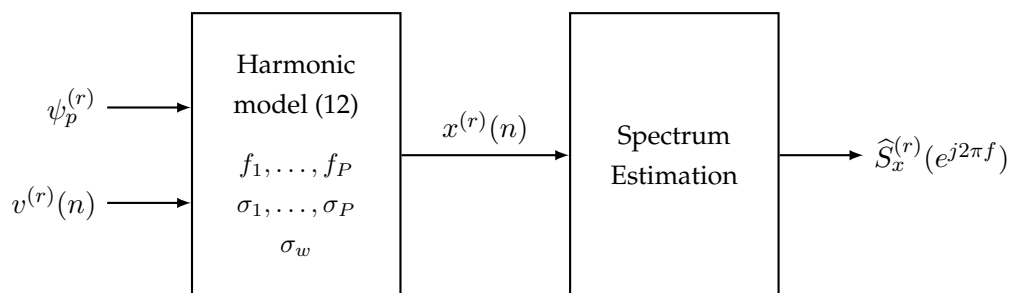


Figure 1: The system diagram for spectrum estimation driven by the *realization*  $v^{(r)}(n)$  in the  $r$ th Monte-Carlo experiment.

- (a) (5 points) Express the output  $y(n)$  into the convolution  $h(n) * x(n)$ . Relate the impulse response  $h(n)$  to the coefficients  $c_0, c_1, \dots, c_{M-1}$ .
- (b) (10 points) Assume that  $x(n)$  is generated from a harmonic model with  $P = 1$ , frequency  $f_1$ , magnitude  $\sigma_1$ , and noise variance  $\sigma_w^2$ . Let  $H(z)$  be the  $z$ -transform of  $h(n)$ . Find a **zero** of  $H(z)$  if  $\mathbf{c} = \mathbf{c}_{\text{MVDR}}$  and  $\sigma_w^2 \rightarrow 0$ .
6. (15 points) Let  $x(n)$  be a WSS random process. The vector form of  $x(n)$  is given by  $\mathbf{x}(n) = [x(n), x(n+1), \dots, x(n+M-1)]^T$ . The correlation matrix  $\mathbf{R}$  is defined as  $\mathbb{E}[\mathbf{x}(n)\mathbf{x}^H(n)]$ . Determine whether the following estimators of  $\mathbf{R}$  are unbiased or not. Why or why not?

- (a) (5 points) The sample correlation matrix:

$$\hat{\mathbf{R}} \triangleq \frac{1}{K} \sum_{k=1}^K \mathbf{x}(k)\mathbf{x}^H(k). \quad (9)$$

- (b) (5 points) The sample correlation matrix with diagonal loading ( $\delta > 0$ ):

$$\hat{\mathbf{R}} \triangleq \frac{1}{K} \sum_{k=1}^K \mathbf{x}(k)\mathbf{x}^H(k) + \delta \mathbf{I}. \quad (10)$$

- (c) (5 points) The exponentially-weighted and diagonally-loaded sample correlation matrix ( $0 < \lambda \leq 1$  and  $\delta > 0$ ):

$$\hat{\mathbf{R}} \triangleq \sum_{k=1}^K \lambda^{K-k} \mathbf{x}(k)\mathbf{x}^H(k) + \delta \cdot \lambda^K \mathbf{I}. \quad (11)$$

7. (Spectrum estimation on harmonic models, 30 points) This problem aims to compare the performance of spectrum estimators through Monte-Carlo experiments. The system model is depicted in Figure 1, where the harmonic model is characterized by

$$x(n) = \sum_{p=1}^P \left( \sigma_p e^{j\psi_p} \right) e^{j2\pi f_p n} + \sigma_w v(n), \quad (12)$$

where the phase  $\psi_p$  and the noise  $v(n)$  are random. We assume that  $\psi_1, \dots, \psi_P$  are independent and uniformly distributed over  $[0, 2\pi]$ . Furthermore,  $v(n)$  is a complex circularly-symmetric white Gaussian noise with zero mean and unit variance.

To conduct Monte-Carlo experiments, please read the file `MSA_HW2_Problem_7.mat` for the realizations  $\psi_p^{(r)}$  and  $v^{(r)}(n)$  in the  $r$ th Monte-Carlo experiment. You will find two matrices in this mat file in the following layout:

$$\mathbf{Psi} \triangleq \begin{bmatrix} \psi_1^{(1)} & \psi_2^{(1)} & \dots & \psi_P^{(1)} \\ \psi_1^{(2)} & \psi_2^{(2)} & \dots & \psi_P^{(2)} \\ \vdots & \vdots & \ddots & \vdots \\ \psi_1^{(R)} & \psi_2^{(R)} & \dots & \psi_P^{(R)} \end{bmatrix}, \quad \mathbf{V} \triangleq \begin{bmatrix} v^{(1)}(0) & v^{(1)}(1) & \dots & v^{(1)}(N-1) \\ v^{(2)}(0) & v^{(2)}(1) & \dots & v^{(2)}(N-1) \\ \vdots & \vdots & \ddots & \vdots \\ v^{(R)}(0) & v^{(R)}(1) & \dots & v^{(R)}(N-1) \end{bmatrix}. \quad (13)$$

where  $R$  is the number of Monte-Carlo experiments. We set the following parameters

$$P = 5, \quad (14)$$

$$f_1 = -0.3, \quad f_2 = -0.2, \quad f_3 = 0.05, \quad f_4 = 0.1, \quad f_5 = 0.4, \quad (15)$$

$$\sigma_1 = 2, \quad \sigma_2 = 1, \quad \sigma_3 = 1.5, \quad \sigma_4 = 1.2, \quad \sigma_5 = 3, \quad (16)$$

$$\sigma_w = 1, \quad (17)$$

$$M = 15. \quad (18)$$

**Save all the plots in Problem 7 to `fig` files. Include these `fig` files in your submission.**

(a) (3 points) Plot the following curves The horizontal axis is time index  $n$ . The vertical axis is the real and imaginary parts of these signals.

- The realization  $x^{(1)}(n)$ .
- The realization  $x^{(2)}(n)$ .
- The realization  $x^{(3)}(n)$ .

(b) (4 points) Let  $\mathbf{R}$  be the true correlation matrix based on (12) and the statistical assumptions of  $\psi_p$  and  $v(n)$ . We have two estimators of  $\mathbf{R}$ :

i. The estimated mean  $\hat{\mathbf{R}}(n)$  over  $R$  realizations at time index  $n$ . Namely,

$$\hat{\mathbf{R}}(n) \triangleq \frac{1}{R} \sum_{r=1}^R \left( \mathbf{x}^{(r)}(n) \right) \left( \mathbf{x}^{(r)}(n) \right)^H. \quad (19)$$

ii. The sample correlation matrix in the  $r$ th Monte-Carlo trial. More specifically,

$$\hat{\mathbf{R}}_L^{(r)} \triangleq \frac{1}{L} \sum_{n=0}^{L-1} \left( \mathbf{x}^{(r)}(n) \right) \left( \mathbf{x}^{(r)}(n) \right)^H. \quad (20)$$

To assess the estimation performance, we define the following error metric

$$\mathcal{E}(\hat{\mathbf{R}}, \mathbf{R}) \triangleq \sqrt{\frac{1}{M^2} \left\| \hat{\mathbf{R}} - \mathbf{R} \right\|_F^2} = \frac{1}{M} \left\| \hat{\mathbf{R}} - \mathbf{R} \right\|_F, \quad (21)$$

where  $\|\cdot\|_F$  denotes the Frobenius norm of a matrix. Next, we plot the following curves. The vertical axis is in the logarithmic scale.

- $\mathcal{E}(\widehat{\mathbf{R}}(n), \mathbf{R})$  over  $n$ , where  $n = 0, 1, \dots, N - M$ .
  - $\mathcal{E}(\widehat{\mathbf{R}}_L^{(1)}, \mathbf{R})$  over  $L$ , where  $L = 1, 2, \dots, N - M + 1$ .
  - $\mathcal{E}(\widehat{\mathbf{R}}_L^{(2)}, \mathbf{R})$  over  $L$ , where  $L = 1, 2, \dots, N - M + 1$ .
  - $\mathcal{E}(\widehat{\mathbf{R}}_L^{(3)}, \mathbf{R})$  over  $L$ , where  $L = 1, 2, \dots, N - M + 1$ .
- (c) (3 points) Let  $\lambda_\ell(\mathbf{A})$  be the eigenvalues of a Hermitian matrix  $\mathbf{A} \in \mathbb{C}^{M \times M}$ . These eigenvalues are sorted in the descending order  $\lambda_1(\mathbf{A}) \geq \lambda_2(\mathbf{A}) \cdots \geq \lambda_M(\mathbf{A})$ . Plot the following curves.
- $\lambda_\ell(\mathbf{R})$  over  $\ell$ , where  $\ell = 1, 2, \dots, M$ .
  - $\lambda_\ell(\widehat{\mathbf{R}}_L^{(1)})$  over  $\ell$ , where  $L = \lfloor 0.5(N - M + 1) \rfloor$  and  $\ell = 1, 2, \dots, M$ .
  - $\lambda_\ell(\widehat{\mathbf{R}}_L^{(1)})$  over  $\ell$ , where  $L = N - M + 1$  and  $\ell = 1, 2, \dots, M$ .
- (d) (10 points) Let  $\widehat{S}_{x,\text{MVDR}}^{(r)}(e^{j2\pi f})$  be the MVDR spectrum associated with  $\widehat{\mathbf{R}}_L^{(r)}$  and  $L = N - M + 1$ . Plot the following curves in one plot for  $-\frac{1}{2} \leq f \leq \frac{1}{2}$ . The vertical axis is in the logarithmic scale.
- The MVDR spectrum  $\widehat{S}_{x,\text{MVDR}}^{(1)}(e^{j2\pi f})$ . Mark the first  $P$  dominant peaks and specify their coordinates.
  - The MVDR spectrum  $\widehat{S}_{x,\text{MVDR}}^{(2)}(e^{j2\pi f})$ . Mark the first  $P$  dominant peaks and specify their coordinates.
  - The MVDR spectrum  $\widehat{S}_{x,\text{MVDR}}^{(3)}(e^{j2\pi f})$ . Mark the first  $P$  dominant peaks and specify their coordinates.
  - The estimated mean of the MVDR spectrum over  $R$  Monte-Carlo trials. More specifically,

$$\widehat{\mathcal{M}}_{\text{MVDR}}(e^{j2\pi f}) \triangleq \frac{1}{R} \sum_{r=1}^R \widehat{S}_{x,\text{MVDR}}^{(r)}(e^{j2\pi f}). \quad (22)$$

- (e) (10 points) Let  $\widehat{P}_{x,\text{MUSIC}}^{(r)}(e^{j2\pi f})$  be the MUSIC pseudospectrum associated with  $\widehat{\mathbf{R}}_L^{(r)}$  and  $L = N - M + 1$ . Plot the following curves in one plot for  $-\frac{1}{2} \leq f \leq \frac{1}{2}$ . The vertical axis is in the logarithmic scale.
- The MUSIC pseudospectrum  $\widehat{P}_{x,\text{MUSIC}}^{(1)}(e^{j2\pi f})$ . Mark the first  $P$  dominant peaks and specify their coordinates.
  - The MUSIC pseudospectrum  $\widehat{P}_{x,\text{MUSIC}}^{(2)}(e^{j2\pi f})$ . Mark the first  $P$  dominant peaks and specify their coordinates.
  - The MUSIC pseudospectrum  $\widehat{P}_{x,\text{MUSIC}}^{(3)}(e^{j2\pi f})$ . Mark the first  $P$  dominant peaks and specify their coordinates.
  - The estimated mean of the MUSIC pseudospectrum over  $R$  Monte-Carlo trials. More specifically,

$$\widehat{\mathcal{M}}_{\text{MUSIC}}(e^{j2\pi f}) \triangleq \frac{1}{R} \sum_{r=1}^R \widehat{P}_{x,\text{MUSIC}}^{(r)}(e^{j2\pi f}). \quad (23)$$

Last updated April 7, 2022.