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**EE 6110: Adaptive Signal Processing**

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**Tutorial #2**

KG/IITM

Note: Questions marked with a “\*” are perhaps of higher difficulty

1. If  $\mathbf{A}^H = -\mathbf{A}$ , (a) show that  $j\mathbf{A}$  is Hermitian ( $j = \sqrt{-1}$ )  
(b)\* show that  $\mathbf{A}$  is unitary, diagonalizable and has pure imaginary eigenvalues.
2. Find  $k, l$ , and  $m$  to make the matrix  $\mathbf{A}$  shown below, a Hermitian matrix

$$\mathbf{A} = \begin{bmatrix} -1 & k & -j \\ 3 - j5 & 0 & m \\ l & 2 + j4 & 2 \end{bmatrix}$$

3. In each of the below, find the unitary matrix  $\mathbf{Q}$  that diagonalizes  $\mathbf{A}$ , i.e.,  $\mathbf{Q}^H \mathbf{A} \mathbf{Q} = \mathbf{\Lambda}$ .

a)

$$\mathbf{A} = \begin{bmatrix} 4 & 1 - j \\ 1 + j & 5 \end{bmatrix}$$

b)

$$\mathbf{A} = \begin{bmatrix} 5 & 0 & 0 \\ 0 & -1 & -1 + j \\ 0 & -1 - j & 0 \end{bmatrix}$$

c)

$$\mathbf{A} = \begin{bmatrix} 2 & \frac{j}{\sqrt{2}} & -\frac{j}{\sqrt{2}} \\ -\frac{j}{\sqrt{2}} & 2 & 0 \\ \frac{j}{\sqrt{2}} & 0 & 2 \end{bmatrix}$$

4. If  $\mathbf{A}$  has  $\lambda_1 = 0$ , and  $\lambda_2 = 5$  corresponding to

$$q_1 = \begin{bmatrix} 1 \\ 2 \end{bmatrix}, \quad q_2 = \begin{bmatrix} 2 \\ -1 \end{bmatrix}$$

Find: (a) trace( $\mathbf{A}$ ), (b) det( $\mathbf{A}$ ), and  
(c) Can you specify  $\mathbf{A}$ ?

- 5.\* Find a real matrix  $\mathbf{A}$  with  $\mathbf{A} + \alpha \mathbf{I}$  invertible for all real  $\alpha$ .

6. Find (i)  $\mathbf{A}^{90}$ , (ii)  $e^{\mathbf{A}}$ , if

$$\mathbf{A} = \begin{bmatrix} 1 & 0 \\ 2 & 1 \end{bmatrix}$$

7.\* Find a 3 x 3 matrix whose rows add up to 1 and show that  $\lambda = 1$  is an eigenvalue of this matrix. What is the corresponding  $\mathbf{q}$ ?

8. Verify if the matrices are unitary; if so, specify their inverses.

a)

$$\mathbf{A} = \begin{bmatrix} -j/\sqrt{2} & j/\sqrt{6} & j/\sqrt{3} \\ 0 & -j/\sqrt{6} & j/\sqrt{3} \\ j/\sqrt{2} & j/\sqrt{6} & j/\sqrt{3} \end{bmatrix}$$

b)

$$\mathbf{A} = \begin{bmatrix} 3/5 & j4/5 \\ -4/5 & j3/5 \end{bmatrix}$$

c)

$$\mathbf{A} = \frac{1}{4} \begin{bmatrix} j + \sqrt{3} & 1 - j\sqrt{3} \\ 1 + j\sqrt{3} & j - \sqrt{3} \end{bmatrix}$$

d)

$$\mathbf{A} = \begin{bmatrix} \frac{1+j}{2} & \frac{-1}{2} & \frac{1}{2} \\ \frac{j}{\sqrt{3}} & \frac{1}{\sqrt{3}} & \frac{-j}{\sqrt{3}} \\ \frac{3+j}{2\sqrt{15}} & \frac{4+3j}{2\sqrt{15}} & \frac{5j}{2\sqrt{15}} \end{bmatrix}$$

9. Consider a recursion where  $\mathbf{x}(0) = [2 \ 0 \ 2]^T$  and  $\mathbf{x}(k+1) = \mathbf{A} \mathbf{x}(k)$

$$\mathbf{A} = \begin{bmatrix} 1/2 & 1/2 & 1/2 \\ 1/4 & 1/2 & 0 \\ 1/4 & 0 & 1/2 \end{bmatrix}$$

Find the (i) eigenvalues and the eigenvectors of the matrix  $\mathbf{A}$ , (ii) the value of  $\mathbf{x}(k+1)$ , (iii) the limiting value  $\mathbf{x}(\infty)$ .

10.\* A WSS process has  $\mathbf{R}_{xx}(0) = 1$ , and  $\mathbf{R}_{xx}(\pm 1) = 0.8$ .

- a) Choose  $\mathbf{R}_{xx}(\pm 2) \geq 0$ , such that the process is deterministic like.
- b) For your choice of  $\mathbf{R}_{xx}(\pm 2)$ , obtain an expression for  $\mathbf{R}_{xx}(k)$  for all  $k$ .

11. Consider a r.p.  $u(n) = 3e^{j4\pi n} + v(n)$ , where  $v(n)$  is a Gaussian white noise process with variance  $\sigma^2 = 4$ .

- a) find the 2 x 2 autocorrelation matrix  $\mathbf{R}$ .
- b) what is the eigenvalue spread of  $\mathbf{R}$ ?
- c) find the expression for  $\mathbf{R}^6$ .

12. We are given random samples  $\{x_1, x_2, \dots, x_N\}$  where each  $x_i$  is i.i.d. with  $N(\mu, \sigma^2)$ . Consider the following estimator for  $\mu$ ,

$$\hat{\mu}(N) = \frac{1}{N+a} \sum_{i=1}^N x_i$$

where  $a \geq 0$ .

- (a) For what value(s) of  $a$  is the above unbiased estimator of  $\mu$ ? (for small sample case)
- (b) For what value(s) of  $a$  is the above an asymptotically unbiased estimator of  $\mu$ ?
- (c)\* Prove that the above is a consistent estimator of  $\mu$ , for all  $a \geq 0$ .

13. Suppose that  $N$  independent observations  $\{x_1, x_2, \dots, x_N\}$  are made of a r.v.  $X$  that is Gaussian; i.e.,

$$p(x_i | \mu, \sigma^2) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp[-(x_i - \mu)^2 / 2\sigma^2]$$

Assuming only  $\mu$  is unknown, derive the Cramer-Rao lower bound (CRLB) of  $E[(\mu - \hat{\mu})^2]$  for an unbiased estimator of  $\mu$ .

14.\* For the observations in Pbm.13, now assume that only variance  $\sigma^2$  is unknown, and derive the CRLB of  $E[(\sigma^2 - \hat{\sigma}^2)^2]$  for an unbiased estimator of  $\sigma^2$ .