

## **Cover Sheet**

Test Duration: 75 minutes.

Open Book but Closed Notes.

Calculators **not** allowed.

This test contains **three** problems.

All work should be done in the blue books provided.

You must show all work for each problem to receive full credit.

Do **not** return this test sheet, just return the blue books.

<b>No.</b>	<b>Topic(s) of Problem</b>	<b>Points</b>
1.	Autoregressive (AR) Spectral Estimation	40
2.	Sum of Sinewaves Spectral Analysis	30
3.	Spectral Characteristics of ARMA Processes	30

# Digital Signal Processing I      Exam 3      30 Oct. 2001

## Problem 1. [30 points]

Consider the autoregressive AR(2) process generated via the difference equation

$$x[n] = \frac{7}{12}x[n-1] - \frac{2}{12}x[n-2] + w[n]$$

where  $w[n]$  is a stationary white noise process with variance  $\sigma_w^2 = 35/6$ . (The value of  $\sigma_w^2$  was chosen so that the autocorrelation values requested in part (a) below are whole numbers.)

- Determine the numerical values of  $r_{xx}[0]$ ,  $r_{xx}[1]$ ,  $r_{xx}[2]$ , where  $r_{xx}[m]$  is the autocorrelation sequence  $r_{xx}[m] = E\{x[n]x[n-m]\}$ .
- Determine a simple closed-form expression for the spectral density for  $x[n]$ ,  $S_{xx}(\omega)$ , which may be expressed as the DTFT of  $r_{xx}[m]$ :

$$S_{xx}(\omega) = \sum_{m=-\infty}^{\infty} r_{xx}[m]e^{-jm\omega}$$

- Consider the first-order predictor

$$\hat{x}[n] = -a_1(1)x[n-1]$$

Determine the numerical value of the optimum predictor coefficient  $a_1(1)$  and the corresponding minimum mean-square error.

- Consider the third-order predictor

$$\hat{x}[n] = -a_3(1)x[n-1] - a_3(2)x[n-2] - a_3(3)x[n-3]$$

Determine the numerical values of the optimum predictor coefficients  $a_3(1)$ ,  $a_3(2)$ , and  $a_3(3)$  and the corresponding minimum mean-square error.

## Problem 2. [30 points]

Consider the discrete-time complex-valued random process defined for all  $n$ :

$$x[n] = e^{j(\omega_1 n + \Theta_1)} + \sqrt{2} e^{j(\omega_2 n + \Theta_2)}$$

where the respective frequencies,  $\omega_1$  and  $\omega_2$ , of the two complex sinewaves are deterministic but unknown constants.  $\Theta_1$  and  $\Theta_2$  are independent random variables with each uniformly distributed over a  $2\pi$  interval. The values of the autocorrelation sequence for  $x[n]$ ,  $r_{xx}[m] = E\{x[n]x^*[n-m]\}$ , for three different lag values are given below.

$$r_{xx}[0] = 3, \quad r_{xx}[1] = -2 + j, \quad r_{xx}[2] = 1$$

- Determine the numerical values of  $\omega_1$  and  $\omega_2$ . **You have to use what you've learned during the parametric spectral analysis portion of this course. You will be given no credit if you simply set up a system of equations to solve based on the form of  $r_{xx}[m] = \sum_{i=1}^p A_i^2 e^{j\omega_i m}$  and solve this nonlinear system of equations.** Part (b) is on top of next page.

(b) Consider a second-order predictor

$$\hat{x}[n] = -a_2(1)x[n-1] - a_2(2)x[n-2]$$

Determine the numerical values of the optimum predictor coefficients  $a_2(1)$  and  $a_2(2)$ , and the numerical value of the corresponding minimum mean-square error.

**Problem 3.** [30 points]

Consider the ARMA(1,1) process generated via the difference equation

$$x[n] = \frac{1}{2}x[n-1] + w[n] + w[n-1]$$

where  $w[n]$  is a stationary white noise process with variance  $\sigma_w^2 = 1$ .

- (a) Determine the numerical values of  $r_{xx}[0]$ ,  $r_{xx}[1]$ ,  $r_{xx}[2]$ , where  $r_{xx}[m]$  is the autocorrelation sequence  $r_{xx}[m] = E\{x[n]x[n-m]\}$ . (Note that  $r_{xx}[m]$  is the inverse DTFT of the spectral density  $S_{xx}(\omega)$  asked for in Part (b) below, but there are at least three different ways you can solve this part of the problem.)
- (b) Determine a simple closed-form expression for the spectral density for  $x[n]$ ,  $S_{xx}(\omega)$ , which may be expressed as the DTFT of  $r_{xx}[m]$ :

$$S_{xx}(\omega) = \sum_{m=-\infty}^{\infty} r_{xx}[m]e^{-jm\omega}$$

(c) Consider the first-order predictor

$$\hat{x}[n] = -a_1(1)x[n-1]$$

Determine the numerical value of the optimum predictor coefficient  $a_1(1)$  and the corresponding minimum mean-square error.