

1. This problem investigates least-squares linear prediction, using the covariance method for determining the optimal coefficients. Consider the signal

$$x(n) = \{2^{-n} + 4^{-n}\}u(n)$$

- a. Calculate an expression for the covariance function of  $x(n)$  in terms of  $k$ ,  $l$ , and  $P$ , for  $0 \leq k \leq P$  and  $0 \leq l \leq P$ .

$$\phi(k, l) = \sum_{n=P}^{\infty} x(n-k)x(n-l)$$

- b. Set up the appropriate equations for determining the optimal coefficients for the second-order predictor ( $P=2$ ) shown below. Then solve for the coefficients  $\alpha_1$  and  $\alpha_2$ . (You may use a calculator/Matlab to determine the covariance values and to solve the system, but be sure to show the equations you are solving.)

$$\hat{x}(n) = \alpha_1 x(n-1) + \alpha_2 x(n-2)$$

- c. Determine the numerical value of the error for this predictor (given below,  $P=2$ ).

$$E = \sum_{n=P}^{\infty} \{x(n) - \hat{x}(n)\}^2$$

- d. Repeat parts (b) and (c) for a first-order predictor ( $P=1$ ). Compare the resulting error to the second-order predictor.
- e. Determine the optimal coefficients for a third-order predictor.  
Hint: This is very easy if you did parts (b) and (c) correctly.

2. This problem uses the autocorrelation method for determining the optimal LPC coefficients. Consider the signal

$$x(n) = \{2^{-n} + 4^{-n}\}u(n)$$

- a. Calculate an expression for the autocorrelation function for  $x(n)$ .

$$R(m) = \sum_{n=-\infty}^{\infty} x(n)x(n-m)$$

- b. Set up the appropriate equations for determining the optimal coefficients for the second-order predictor (P=2) shown below. Then solve for the coefficients  $\alpha_1$  and  $\alpha_2$ . (You may use a calculator/Matlab to determine the autocorrelation values and to solve the system, but be sure to show the equations you are solving.)

$$\hat{x}(n) = \alpha_1 x(n-1) + \alpha_2 x(n-2)$$

- c. Determine the numerical value of the error for this predictor (given below). NOTE that this expression is different from problem (1).

$$E = \sum_{n=-\infty}^{\infty} \{x(n) - \hat{x}(n)\}^2$$

- d. Compare the coefficients and error to the P=2 results from the covariance method of problem (1).

3. Define the complex cepstrum  $\hat{x}(n)$  of a DT signal  $x(n)$  as the inverse DTFT of the log of the DTFT of  $x(n)$ :

$$\hat{x}(n) = \frac{1}{2\pi} \int_{-\pi}^{\pi} \log[X(e^{j\omega})] e^{j\omega n} d\omega$$

Consider an all-pole model of the vocal tract given by the following transfer function:

$$V(z) = \prod_{k=1}^Q \frac{1}{(1 - p_k z^{-1})(1 - p_k^* z^{-1})}$$

where  $p_k = r_k e^{j\theta_k}$  and  $0 < r_k < 1$ , for  $k=1, \dots, Q$ . Let  $v(n)$  be the inverse Z-transform of  $V(z)$  (i.e.  $v(n)$  is the impulse response of the vocal tract). Show that the cepstrum of  $v(n)$  may be expressed as

$$\hat{v}(n) = 2 \sum_{k=1}^Q \frac{(r_k)^n}{n} \cos(\theta_k n) u(n-1)$$

Hint: You may want to use the following DTFT relationship:

$$n x(n) \stackrel{DTFT}{\Leftrightarrow} j \frac{d}{d\omega} X(e^{j\omega})$$