

Tuesday 30 April 2013 2 to 3.30

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Module 4F7

DIGITAL FILTERS AND SPECTRUM ESTIMATION

*Answer not more than **three** questions.*

*All questions carry the same number of marks.*

*The **approximate** percentage of marks allocated to each part of a question is indicated in the right margin.*

*There are no attachments.*

STATIONERY REQUIREMENTS

Single-sided script paper

SPECIAL REQUIREMENTS

Engineering Data Book

CUED approved calculator allowed

**You may not start to read the questions  
printed on the subsequent pages of this  
question paper until instructed that you  
may do so by the Invigilator**

1 Let

$$x(n) = \sum_{i=1}^L a_i x(n-i) + w(n)$$

where  $\{w(n)\}$  is an independent and identically distributed zero-mean unit-variance noise sequence.

- (a) Formulate the Wiener filtering problem whose solution is  $(a_1, a_2, \dots, a_L)$ . [20%]
- (b) Write down the Steepest Descent algorithm for estimating  $(a_1, a_2, \dots, a_L)$ . [15%]
- (c) Write down the Least Mean Square (LMS) algorithm for estimating  $(a_1, a_2, \dots, a_L)$ . [10%]
- (d) Rather than  $x(n)$  itself, only noisy observations  $u(n) = x(n) + v(n)$  are available where  $\{v(n)\}$  is an independent and identically distributed zero-mean noise sequence, independent of  $\{w(n)\}$ , with variance  $\sigma_v^2$ .
- (i) Write down the LMS algorithm for estimating  $(a_1, a_2, \dots, a_L)$  and characterise the limit of the expected value of the LMS estimates. [35%]
- (ii) Assuming knowledge of  $\sigma_v^2$ , modify your LMS algorithm so that the expected value of the LMS estimates converge to  $(a_1, a_2, \dots, a_L)$ . You must also prove convergence. (Hint: In your LMS algorithm, use an estimate of the autocorrelation of  $\{x(n)\}$  instead.) [20%]

2 (a) Describe, with the definition of the relevant signals, the four main applications of adaptive filters. [30%]

(b) Figure 1 depicts a finite impulse response (FIR) identification problem. The coefficients of the FIR filter,  $(\beta_0, \beta_1, \dots, \beta_{L-1})$ , are to be estimated using an input sequence  $\{u(n)\}_{n \geq 0}$ . However, only noisy measurements  $z(n) = y(n) + v(n)$  of the FIR filter output  $y(n)$  are available where  $\{v(n)\}$  is an independent and identically distributed zero-mean noise sequence with variance  $\sigma_v^2$ .

(i) Describe in detail the Recursive Least Squares (RLS) method for adaptively estimating the coefficients of the FIR filter and contrast its performance with the Least Mean Square method. [40%]

(ii) Let  $L = 3$  and assume knowledge of  $\beta_0 = \beta_1 = 1$ . Compute the RLS estimate of  $\beta_2$  at time  $n$ . State the limit of the RLS estimate. [30%]

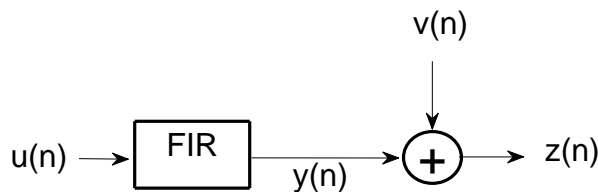


Fig. 1

3 Consider the following ARMA(2,2) model

$$x_n = a_2 x_{n-2} + b_0 w_n + b_1 w_{n-1}$$

where  $\{w_n\}$  is a sequence of independent Gaussian random variables with zero mean and unit variance. Let  $R_{XX}[k] = E \{x_{n+k} x_n\}$ .

- (a) Derive 3 simultaneous equations for  $R_{XX}[0]$ ,  $R_{XX}[1]$  and  $R_{XX}[2]$ . [25%]
- (b) Consider the even index variables  $(x_0, x_2, \dots, x_N)$  where  $N \geq 2$  is even.
- (i) Write down the expression for the joint probability density of  $(x_2, x_4, \dots, x_N)$  given  $x_0$ . [20%]
- (ii) Solve for the Maximum Likelihood estimates of  $a_2$  and  $b_0^2 + b_1^2$  using just  $(x_0, x_2, \dots, x_N)$ . [35%]
- (c) You are given  $(x_0, x_1, x_2, \dots, x_N)$  from which you form an estimate of  $R_{XX}[1]$ . Describe in detail how estimates for  $b_0$  and  $b_1$  may now be obtained. [20%]

4 (a) For a scalar random process  $\{x_n\}$ , let  $R_{XX}[k] = E\{x_{n+k}x_n\}$ . For each autocorrelation function  $R_{XX}$  below, give the equation of the corresponding ARMA process and verify your answer.

(i)  $R_{XX}[2k] = \alpha^k$  and  $R_{XX}[2k+1] = 0$  where  $|\alpha| < 1$ . [20%]

(ii)  $R_{XX}[k] = 10 - |k|$  for  $|k| \leq 10$ , and  $R_{XX}[k] = 0$  otherwise. [20%]

(b) Let  $y_n = \beta y_{n-1} + x_n$ ,  $|\beta| < 1$ , where  $x_n$  is the AR(1) process

$$x_n = \alpha x_{n-1} + w_n$$

with  $|\alpha| < 1$ , and  $w_n$  is a sequence of independent random variables with zero mean and unit variance.

(i) Obtain the expression for  $R_{XX}[k]$ . [15%]

(ii) Write  $y_n$  as a convolution of the sequence  $\{x_k\}$  and then derive the expression for  $R_{YX}[n+k, n] = E\{y_{n+k}x_n\}$ . [15%]

(iii) Obtain the power spectrum of  $\{y_n\}$  in terms of the power spectrum of  $\{x_n\}$ . (Hint: For example, first obtain an expression for  $R_{YY}[k]$  in terms of  $R_{YY}[k+1]$  and  $R_{YX}[n+k, n]$ .) [30%]

**END OF PAPER**