Department of Electrical & Computer Engineering University of California, Santa Barbara ECE 245 Spring 2011 Shynk H.O. #14

## HOMEWORK #5

Due Friday, May 6, 2011 (5:00 p.m.)

**Reading:** Chapter 8 (8.1-8.8)

## **Problems:**

- 1. Chapter 6: Problem 1
- 2. Chapter 6: Problem 7
- 3. Verify equations (42) and (46) in the paper by Horowitz and Senne.
- 4. Consider the modified LMS algorithm

$$W(n+1) = W(n) + 2\alpha R^{-1}X(n)e(n)$$

where R is the autocorrelation matrix of the input signal X(n).

- (a) Find the range of values for  $\alpha$  such that W(n) converges in the mean.
- (b) Specify the value of  $\alpha$  which results in one-step convergence in the mean.
- (c) Derive an equation for this algorithm similar to (37) in the paper mentioned above, and indicate how the new equation is different from (37).
- 5. Consider a system-identification application where the desired response d(n) is obtained as follows (using mixed time-domain and z-domain notation):

$$d(n) = G(z)x(n) + v(n)$$

where the input signal x(n) is a zero-mean random sequence with autocorrelation matrix R, and G(z) is the system to be identified. The additive white-noise sequence v(n) has zero mean and variance  $\sigma_v^2$ , and is uncorrelated with x(n). Assume that the LMS algorithm with M coefficients is used to identify

$$G(z) = 1 - 0.5z^{-1} + 0.2z^{-2}$$

where

$$R = \left[ \begin{array}{rrrr} 1 & 0.3 & 0 \\ 0.3 & 1 & 0.3 \\ 0 & 0.3 & 1 \end{array} \right],$$

 $\sigma_v^2 = 0.02$ , and M = 3.

- (a) Determine  $\xi_{\min}$  for the adaptive filter, and find the Wiener weight vector.
- (b) For convergence in the mean, find the stable range of  $\alpha$ .
- (c) Repeat part (b) for convergence in the mean square (using the results of Horowitz and Senne).
- (d) Plot D(z) for real z (from the paper mentioned above) showing all M roots for a value of  $\alpha$  that satisfies (b) but not (c).
- 6. (a) Verify that the following two LMS stability conditions are equivalent:

$$\begin{array}{ll} 0 < \mu < 1/3 \lambda_{\max} & \mbox{ and } & \eta(\mu) < 1 \\ 0 < \mu < 1/2 \lambda_{\max} & \mbox{ and } & \eta(\mu) < 1 \end{array}$$

where

$$\eta(\mu) = \sum_{k=0}^{M-1} \frac{\mu \lambda_k}{1 - 2\mu \lambda_k}.$$

(b) Find examples for M = 3 where (i)  $\eta(\mu) > 1$ ,  $\mu < 1/2\lambda_{\text{max}}$ , and (ii)  $\eta(\mu) < 1$ ,  $\mu > 1/2\lambda_{\text{max}}$ .