Department of Electrical & Computer Engineering University of California, Santa Barbara ECE 245 Spring 2011 Shynk H.O. #7

HOMEWORK #3

Due Friday, April 22, 2011 (5:00 p.m.)

Reading: Chapter 5 (5.1-5.7)

Problems:

- 1. Chapter 2: Problem 15
- 2. Chapter 4: Problem 5
- 3. Chapter 4: Problem 12
- 4. Suppose that we have a tapped-delay line with mean-square error (MSE)

$$\xi = 2w_0^2 + 2w_1^2 + 2w_0w_1 - 14w_0 - 16w_1 + 42$$

where $\phi_{xx}(0) = 2$ and $\phi_{xx}(1) = 1$.

- (a) Assume that the gradient estimate is based on 100 observations of the error at each perturbed weight setting. Find the covariance matrix of the gradient estimate, assuming that e(n) has a Gaussian distribution with zero mean and variance σ^2 .
- (b) Determine the weight-vector covariance matrix. Assume that the steepest-descent (SD) algorithm is used with step-size parameter α equal to one half its maximum stable value, and that there are 100 error observations for each weight setting. Find the excess mean-square error and the misadjustment M.
- 5. Consider using the SD algorithm to solve the Wiener-Hopf equation.
 - (a) Show that the kth component of the weight vector can be written in rotated (but not translated) coordinates as follows:

$$W_k'(n+1) = (1 - 2\alpha\lambda_k)W_k'(n) + 2\alpha P_k'$$

where P' is the rotated cross-correlation vector.

- (b) Using the one-sided z-transform, find a closed-form expression for $W'_k(n)$ in terms of the initial weight $W'_k(0)$.
- (c) Using the result in part (b), find the limit as $n \to \infty$. Is your answer consistent with that obtained using the rotated and translated weight vector V'?
- 6. Derive a bound on α for the SD algorithm such that the rotated and translated weights V'(n) converge in a monotonic manner to their optimal values (i.e., there is no overshoot or oscillation as the uncoupled weights evolve). How does your answer compare to that obtained for Newton's method (NM)?
- 7. Suppose that e(n) is a Gaussian random process with zero mean and variance σ^2 . Let

$$\hat{\xi} = \frac{1}{N} \sum_{k=0}^{N-1} e^2(k)$$

and assume that the $\{e(k)\}$ are independent. Note that $E[\hat{\xi}] = \sigma^2 = \xi$, i.e., $\hat{\xi}$ is an unbiased estimate of the MSE. Derive an expression for the variance of $\hat{\xi}$.