Reading: Chapter 5 (5.1–5.7)

Problems:

1. Chapter 2: Problem 13
2. Chapter 4: Problem 5
3. Chapter 4: Problem 9
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 \( \sigma^2 \).
   (b) Determine the weight-vector covariance matrix. Assume that the steepest-descent (SD)
       algorithm is used with step-size parameter \( \alpha \) 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 \( k \)th component of the weight vector can be written in rotated (but not
       translated) coordinates as follows:
       \[ W'_k(n + 1) = \left( 1 - 2\alpha\lambda_k \right) W'_k(n) + 2\alpha P'_k \]
       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 $U$?

6. Derive a bound on $\alpha$ 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 $\sigma^2$. 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 $\hat{\xi}$. 