

- 9.8** Suppose the mean square error is approximated using a running average filter of order $M - 1$ as follows.

$$\epsilon(w) \approx \frac{1}{M} \sum_{i=0}^{M-1} e^2(k-i)$$

- (a) Find an expression for the gradient vector $\nabla \epsilon(w)$ using this approximation for the mean square error.
- (b) Using the steepest descent method and the results from part (a), find a weight update formula.
- (c) How many floating point multiplications (FLOPs) are required per iteration to update the weight vector? You can assume that 2μ is computed ahead of time.
- (d) Verify that when $M = 1$ the weight update formula reduces to the LMS method.

Solution

- (a) Using (9.2.3) and (9.2.4), the partial derivative of this approximation to the mean square error with respect to w_i is

$$\begin{aligned} \frac{\partial \epsilon(w)}{\partial w_i} &= \left(\frac{\partial}{\partial w_i} \right) \frac{1}{M} \sum_{q=0}^{M-1} e^2(k-q) \\ &= \frac{1}{M} \sum_{q=0}^{M-1} 2e(k-q) \frac{\partial e(k-q)}{\partial w_i} \\ &= \frac{2}{M} \sum_{q=0}^{M-1} e(k-q) \left(\frac{\partial}{\partial w_i} \right) [d(k-q) - y(k-q)] \\ &= \frac{-2}{M} \sum_{q=0}^{M-1} e(k-q) \left(\frac{\partial}{\partial w_i} \right) y(k-q) \\ &= \frac{-2}{M} \sum_{q=0}^{M-1} e(k-q) \left(\frac{\partial}{\partial w_i} \right) w^T \theta(k-q) \\ &= \frac{-2}{M} \sum_{q=0}^{M-1} e(k-q) \theta_i(k-q) \quad , \quad 0 \leq i \leq m \end{aligned}$$

Thus the gradient vector is

$$\nabla \epsilon(w) = \frac{-2}{M} \sum_{i=0}^{M-1} e(k-i) \theta(k-i)$$

- (b) Using (9.3.3) and the results from part (a), the weight update formula using the steepest descent method is

$$\begin{aligned} w(k+1) &= w(k) - \mu \nabla \epsilon[w(k)] \\ &= w(k) + \frac{2\mu}{M} \sum_{i=0}^{M-1} e(k-i) \theta(k-i) \end{aligned}$$

- (c) Since $\theta(k-i)$ is a vector of length $m+1$, the number of FLOPs per iteration to update the weight vector is

$$n_w = (m+1)(M+1)$$

- (d) Starting with the answer to part (b) and setting $M=1$ yields

$$\begin{aligned} w(k+1) &= w(k) + \frac{2\mu}{M} \sum_{i=0}^{M-1} e(k-i) \theta(k-i) \\ &= w(k) + 2\mu e(k) \theta(k) \quad \checkmark \end{aligned}$$