

E9 211 Adaptive Signal Processing
4 December 2019, 14:00–17:00, Final Exam

This exam has three questions (50 points).

One A4 cheat sheet is allowed. No other materials will be allowed.

You need to submit **Homework 3 by the deadline 10 December 2019** and should have already submitted Homeworks 1 and 2 to pass this course. If you have not yet submitted Homeworks 1 and 2, do it before 8 December 2019.

Question 1 (15 points)

Consider a filter \mathbf{w} , which is applied to \mathbf{x}_k to estimate the desired signal d_k as

$$\hat{d}_k = \mathbf{w}^H \mathbf{x}_k.$$

Assume that the desired signal has unit power, i.e., $E\{|d_k|^2\} = 1$. In this problem, we will find the filter by optimizing the following cost functions:

$$J(\mathbf{w}) = E\{|d_k - \mathbf{w}^H \mathbf{x}_k|^2\}$$

and

$$J'(\mathbf{w}) = E\{|d_k - \mathbf{w}^H \mathbf{x}_k|^2\} + \beta \|\mathbf{w}\|_2^2,$$

where $\beta > 0$. To answer this question, use the following definitions $E\{\mathbf{x}_k \mathbf{x}_k^H\} = \mathbf{R}_x$ and $E\{\mathbf{x}_k \bar{d}_k\} = \mathbf{r}_{xd}$.

- (2 pts) (a) Find the optimal filter that minimizes the cost $J(\mathbf{w})$. Compute the resulting minimum cost $J(\mathbf{w}_{\text{opt}})$.
- (4 pts) (b) Find the optimal filter that minimizes the cost $J'(\mathbf{w})$. Compute the resulting minimum cost $J'(\mathbf{w}_{\text{opt}})$.
- (2 pts) (c) Show that $J'(\mathbf{w}_{\text{opt}}) > J(\mathbf{w}_{\text{opt}})$.
- (2 pts) (d) Give the steepest-descent update equations to find the minimizer of $J(\mathbf{w})$. What is the maximum step size μ that can be used so that the steepest-descent algorithm converges?
- (4 pts) (e) Derive the steepest-descent algorithm for minimizing $J'(\mathbf{w})$ and determine the condition on the step size so that the iterations converge.
- (1 pts) (f) When is the algorithm in (e) more useful than the algorithm in (d)?

Question 2 (15 points)

Let us consider the problem of estimating an unknown constant x given measurements that are corrupted by uncorrelated, zero mean noise $v(n)$ that has a variance σ_v^2 . The measurement equation is

$$y(n) = x(n) + v(n)$$

Since the value of x does not change with time n , we have

$$x(n) = x(n-1).$$

- (2 pts) (a) Assume that we are at time step N and gathered all the measurements $\{y(1), y(2), \dots, y(N)\}$. Compute the least squares solution for x .
- (5 pts) (b) Give the recursive least squares (RLS) update equations assuming that the observations arrive sequentially.
- (5 pts) (c) Derive now the Kalman filter update equations.
- (3 pts) (d) Let us denote $P(n-1|n-1) = P(n-1)$. Show that

$$P(n) = \frac{P(0)\sigma_v^2}{nP(0) + \sigma_v^2}.$$

Question 3 (20 points)

In many signal processing applications, it is important to design a filter with *linear phase* without which the phase distortion introduced by the filter might severely degrade the estimated signal. Therefore, we would like to design an adaptive linear phase filter $\mathbf{w}_k = [w_k(0), w_k(1), \dots, w_k(P)]^T$ whose weights at each time k satisfy the following symmetry constraint

$$w_k(n) = w_k(P-n), \quad n = 0, 1, \dots, P.$$

- (3 pts) (a) Formulate the symmetry constraint as a linear equality constraint $\mathbf{C}^H \mathbf{w}_k = \mathbf{f}$. What will be \mathbf{C} and \mathbf{f} for the special case with $P = 2$?
- (6 pts) (b) Find the optimal solution to the constrained optimization problem

$$\text{minimize } E\{|d_k - \mathbf{w}^H \mathbf{x}_k|^2\} \quad \text{subject to } \mathbf{c}^H \mathbf{w} = f.$$

Notice that \mathbf{c} is a vector and f is a scalar.

- (6 pts) (c) Derive a linearly constrained adaptive LMS filter \mathbf{w}_k that approximates the optimal solution in (b) by minimizing the instantaneous error as

$$\text{minimize } |d_k - \mathbf{w}_k^H \mathbf{x}_k|^2 \quad \text{subject to } \mathbf{c}^H \mathbf{w}_k = f.$$

Hint: Use the extended cost function with a Lagrange multiplier and enforce the condition that each successive weight vector, including the initial condition, satisfies the linear equality constraint.

- (5 pts) (d) It is possible to eliminate the equality constraint by modifying the input signal \mathbf{x}_k to \mathbf{z}_k . The vector \mathbf{z}_k will be of smaller length as compared to \mathbf{x}_k . For $P = 2$, what is \mathbf{z}_k ? Give the standard (unconstrained) LMS update equations for computing the adaptive filter $\tilde{\mathbf{w}}_k$ that will be applied on \mathbf{z}_k to estimate the desired signal as $\tilde{\mathbf{w}}_k^H \mathbf{z}_k$.