

(c) Determine the sequence(s) for which the worst-case ISI occurs.

(d) Show that for an arbitrary choice of N and Δ the optimum solution \mathbf{c}_{opt} in the sense of a minimum mean-squared error (MinMSE) is the $(\Delta + 1)$ -th column of the pseudo-inverse of the convolution matrix of \mathbf{h} , i.e., of

$$(\mathbf{H}^T \mathbf{H})^{-1} \mathbf{H}^T$$

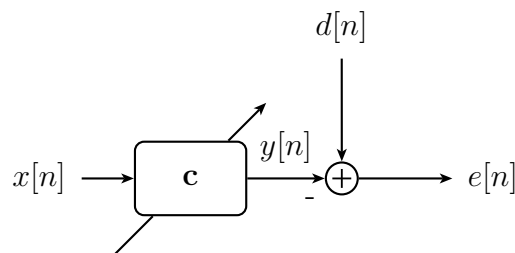
Make sure that you write down the matrix/vector dimensions explicitly!

(e) Now choose $N = 2$ and $\Delta = 0$ and determine the optimum filter coefficient vector \mathbf{c}_{opt} . Calculate the noise gain of the equalizer. Can this equalizer open the channel's eye? If not, write down again the input sequence(s) for which an error occurs.

(f) Repeat task (e) for $\Delta = 1$ and compare the results to those of (e). Which delay Δ is preferable? Justify your answer!

Problem 2 (34 Points)

Consider the following linear filtering problem:



The auto-correlation sequence of $x[n]$ and the cross-correlations between $x[n-k]$ and $d[n]$ are assumed to be known (we can build the auto-correlation matrix \mathbf{R}_{xx} and the cross-correlation vector \mathbf{p}). The following adaptation rule (*coefficient-leakage gradient search*) is used to adapt the coefficient vector $\mathbf{c}[n]$:

$$\mathbf{c}[n] = (1 - \mu\alpha)\mathbf{c}[n-1] + \mu(\mathbf{p} - \mathbf{R}_{xx}\mathbf{c}[n-1])$$

where α is the leakage parameter ($0 < \alpha \ll 1$) and μ is the step-size parameter.

(a) Assume convergence. Where does this algorithm converge to?

(b) Transform the given adaptation rule in a way such that it adapts the misalignment vector (defined as $\mathbf{v}[n] = \mathbf{c}[n] - \mathbf{c}_{\infty}$). Hint: substitute for \mathbf{p} using the result from (a).

(c) Decouple the adaptation rule of (b) into a set of scalar adaptation expressions by using a unitary transform $\mathbf{q}[n] = \mathbf{Q}^H \mathbf{v}[n]$.

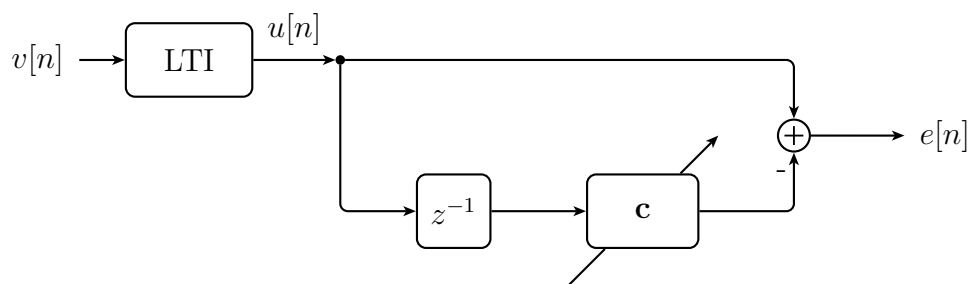
(d) Write the decoupled equation from (c) as individual exponential sequences and derive a condition on μ to ensure convergence towards \mathbf{c}_{∞} , i.e., specify the range $\mu_{\min} < \mu < \mu_{\max}$ (assume α to be given).

(e) Assume we use a fixed step-size parameter μ . Compute the worst-case convergence time constant τ_{max} .

(f) Now set $\mathbf{R}_{xx} = \mathbf{I}$, $\mu = 0.01$, and $\alpha = 0$. How many iterations are necessary until the largest component of the misalignment vector $\mathbf{v}[n]$ reduced to only 5% of its initial value?

Problem 3 (33 Points)

A prediction-error filter can be considered as a blind equalizer for AR processes.



The first 3 samples of the auto-correlation sequence of $u[n]$ are given as $r_{uu}[0] = 1.0$, $r_{uu}[1] = -0.5 = -\frac{5}{9}$, and $r_{uu}[2] = \frac{25}{81} \approx 0.309$ where the bar denotes repeating decimals.

(a) Derive the MSE-optimal coefficient vector \mathbf{c}_{opt} for a general predictor order N . **Hint:** You may use the Wiener-Hopf solution, but make sure that you adapt it properly to the signal model relevant in this example.

(b) For a predictor with $N = 1$ coefficient, find the MSE-optimal coefficient of the adaptive filter.

(c) For a predictor with $N = 2$ coefficients, find the MSE-optimal coefficient vector \mathbf{c} of the adaptive filter.

(d) Given your state of information, what is the order of the AR process? Write down the process-generator difference equation and calculate the variance σ_v^2 of the white-noise input $v[n]$.

(e) Next, assume that the coefficient vector \mathbf{c} ($N = 2$) is updated constantly, such that it is MSE-optimal. Assume now that the input $v[n]$ is suddenly switched from a white to a non-white signal. How would the coefficients change?