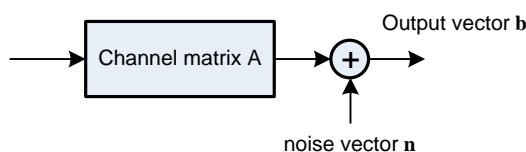


2021 Spring VLSI DSP Homework Assignment #1

Due date: 2021/3/30

Q1. Least Square Problem

Create an arbitrary matrix $\mathbf{A}_{6 \times 4}$ and an arbitrary vector $\mathbf{x}_{4 \times 1}$. The matrix \mathbf{A} should have full column rank, which is 4, and with entries $|a_{i,j}| \leq 10$. The vector \mathbf{x} has all its entry value $|x_i| \leq 4$. Let $\mathbf{n}_{6 \times 1}$ be an arbitrary noise vector with all its entry value of ± 1 . Suppose $\mathbf{A}_{6 \times 4}$ is the channel matrix, $\mathbf{x}_{4 \times 1}$ is the original signal, and $\mathbf{b}_{6 \times 1}$ is the observed output with additive noise $\mathbf{n}_{6 \times 1}$.



We therefore encounter a LS problem to solve $\mathbf{x}_{4 \times 1}$ subject to $\mathbf{A}_{6 \times 4} \cdot \mathbf{x}_{4 \times 1} = \mathbf{b}_{6 \times 1}$.

- Calculate vector $\mathbf{b}_{6 \times 1}$ first use the \mathbf{A} , \mathbf{x} , \mathbf{n} you create. Then calculate the least square estimate $\hat{\mathbf{x}}_{4 \times 1}$ using the pseudo inverse matrix scheme $\hat{\mathbf{x}} = (\mathbf{A}^T \cdot \mathbf{A})^{-1} \cdot \mathbf{A}^T \cdot \mathbf{b}$ and determine its norm 2 value of error vector $(\mathbf{x}_{4 \times 1} - \hat{\mathbf{x}}_{4 \times 1})$.
- Repeat problem a) by using the MatLab QR function, i.e. $[\mathbf{Q}, \mathbf{R}] = \text{qr}[\mathbf{A}]$. Note that the returned \mathbf{Q} is a 6×6 unitary matrix and \mathbf{R} is a 6×4 upper triangular matrix with the last two rows nullified. Calculate $\tilde{\mathbf{b}}_{6 \times 1} = \mathbf{Q}^T \cdot \mathbf{b}$ first and set $\hat{\mathbf{b}}_{4 \times 1}$ as the upper part of $\tilde{\mathbf{b}}_{6 \times 1}$.

Then obtain the LS estimate as $\hat{\mathbf{x}}_{4 \times 1} = \mathbf{R}_{4 \times 4}^{-1} \cdot \hat{\mathbf{b}}_{4 \times 1}$

- Repeat problem b) except now using the Givens rotations to perform the QR factorization
- Compare the LS solutions obtained from the three approaches. Are they identical? Are the norm 2 values of the three error vectors the same?

Q2. LMS filter design

For a least mean square (LMS) adaptive filter, assume the filter is of the form finite impulse response (FIR) and 15-tap long (i.e., with 15 coefficients $b_0 \sim b_{14}$ for $x(n) \sim x(n-14)$). Given an input signal consisting of 2 frequency components

$$s(n) = \sin(2\pi n/12) + \cos(2\pi n/4)$$

develop an adaptive low pass filter design

Set the target as a low pass filter to remove the high frequency component $\cos(2\pi n/4)$ and use $\sin(2\pi n/12)$ as the desired (or training) signal for LMS adaptation. Assume the step

size μ is 2^{-2} .

- write a Matlab code to simulate the LMS based adaptive filtering. Calculate the RMS (root mean square) value of the latest 16 prediction errors (i.e., $r = \sqrt{(e^2(n) + e^2(n-1) + \dots + e^2(n-15))/16}$) and the adaptation is considered being converged if this value is less than 10% of RMS (root mean square) value of the desired signal, which equals $0.1/\sqrt{2}$.
- Show the plot of “r” versus “n” and indicate when the filter converges, i.e. how many training samples are required
- Show the plot of filter coefficients $b_i(n)$, for $i = 0 \sim 14$, versus “n” and see if the values of filter coefficients remain mostly unchanged after convergence
- Apply a 64-point FFT to the impulse response of the converged filter and verify the filter is indeed a low pass one. Note that the input vector to the 64-point FFT is $(b_0, b_1, \dots, b_{14}, 0, 0, \dots, 0)$ with 49 trailing zeros.
- Change the step size μ to 2^{-4} and see how the behavior of the adaptive filter changes.
- Conduct simulation with a sufficiently large number of samples to see how small the value of “r” can be (the convergence bias)