

1. *Expectation and Ergodicity*: For the 3 difference equations given below, estimate the $M \times M$ autocorrelation matrix \mathbf{R}_{uu} for both $M=2$ and $M=3$ as follows:

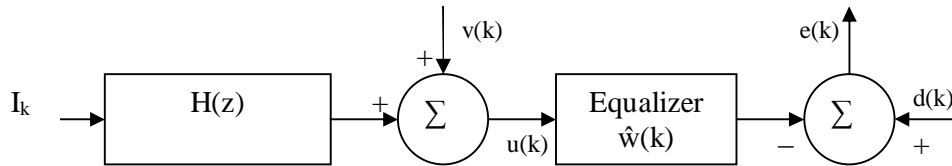
- (i) Calculate the statistical \mathbf{R}_{uu}
- (ii) Assuming ergodicity, numerically estimate \mathbf{R}_{uu} . Consider time averages obtained using 10, 100 and 1000 sample-functions (Monte-Carlo trials). *Hint*: for AR process, to get WSS, drop the first 500-1000 data samples.

Here, $v(n)$ is AWGN with variance as specified below.

- (a) $u(n) = \cos(2\pi n/50) + v(n); \quad \sigma_v^2=0.1$
- (b) $u(n) = 0.8 v(n) - 0.2 v(n-1); \quad \sigma_v^2=1$
- (c) $u(n) = 0.9 u(n-1) + v(n); \quad \sigma_v^2=0.5$

Bonus question: Set up the Yule-Walker equations (read up from Haykin) for the 3 cases, and verify that $\mathbf{w}=\mathbf{R}_{uu}^{-1}\mathbf{r}$ makes sense

2. *Wiener & Least Squares filtering*: In the figure below, the input $\{I_k\}$ is i.i.d. with $E[I_k^2]=1$, following a uniform pdf (i.e., $P(I_k=+1)=1/2=P(I_k=-1)$), and the noise $v(k)$ is AWGN with variance σ_v^2



The bipolar symbols $\{I_k\}$ suffer inter-symbol interference (ISI) from the channel $H(z) = 1 - z^{-1} + 0.5 z^{-2}$, and this ISI is compensated by a M^{th} order equalizer, $\hat{\mathbf{w}}$ (or $\hat{\mathbf{w}}(k)$), defined using the LMSE criterion (Wiener filter), or, using a Least Squares criterion. The desired signal is given by $d(k)$, where $d(k)=I_{k-\Delta}$, with $\Delta \geq 0$ being an integer.

- (a) Write a program to calculate the Wiener-Hopf solution $\hat{\mathbf{w}}=\mathbf{R}^{-1}\mathbf{p}$ which should work for any M and Δ . (Assume $H(z)$ and σ_v^2 are known). For the given $H(z)$, the program should be able to calculate $\hat{\mathbf{w}}$ and J_{\min} for any σ_v^2, M & Δ .
- (b) To study the effect of window length on the “quality” of \mathbf{R} , simulate random data $\{I_k\}$ and $\{v(k)\}$, $k=1 \dots N$, and find the LS estimate $\hat{\mathbf{w}}_{LS}(n)$ when $\Delta=0$, $\sigma_v^2=0.05$, $M=7$, and calculate the least sum of error squares J_{\min}^{LS} for the following values of N : (i) $N=10$, (ii) $N=100$, (iii) $N=200$, and (iv) $N=2000$.
- (c) With $N=1000$, for each of the below situations, find J_{\min} and J_{\min}^{LS} and tabulate them. Comment on your results.

M	Δ	σ_v^2	J_{\min}	J_{\min}^{LS}
3	1	0.01		
10	1	0.01		
15	1	0.01		
10	2	0.01		
10	4	0.01		
10	8	0.01		
10	3	0.01		
10	3	0.05		
10	3	0.10		

Hint: For J_{\min}^{LS} get $\hat{\mathbf{w}}_{LS}$ and substitute this in $\sigma_d^2 - \mathbf{p}^T \hat{\mathbf{w}}_{LS}$