## Department of Electrical Engineering Indian Institute of Technology, Madras

Due before Friday 5pm Dec. 03, 2021

## **EE 6110: Adaptive Signal Processing**

November 16, 2021	Assignment #2	Marks: 20
		Marks. 20

Consider the adaptive channel equalisation model as shown in the figure below (see also pg. 224 in the E-copy of Prof.Ali Sayed's "Adaptive Filter Theory" for a similar problem). Here, the independent, uniformly distributed data symbols I(k) entering the channel F(z) are drawn from a 16-QAM alphabet, and the transmit signal power  $E[|I(k)|^2] = \sigma_1^2 = 1$ . The 3-tap channel is specified by  $H(z) = 1 - 0.8z^{-1} + 0.5 z^{-2}$  and the additive Gaussian noise component v(k) is zero mean with variance  $\sigma_v^2 = 0.02$ . The linear equaliser has order M=14, and the desired response d(k)=I(k- $\Delta$ ), where the decoding delay  $\Delta$ =5.



(a) Determine the LMSE (Wiener) solution  $\mathbf{w}_{opt}$  for this choice of M,  $\Delta$ , and  $\sigma_v^2$ . What is the corresponding  $J_{min}$ ?

(b) Simulate the Least-Mean Squares (LMS) algorithm based adaptive equaliser with  $\mu$ =0.10 being the gain constant. Plot the error convergence curve by averaging e<sup>2</sup>(k) over 25 Monte-Carlo runs, where the average squared error is defined by  $\xi(k) = (1/25) \sum_{i=1}^{25} e_i^2(k)$ . Plot  $10\log_{10}(\xi(k))$  versus k, for k=1,2...2000. What is the simulated and theoretical excess MSE (EMSE) that you get?

(c) For the same channel conditions and M and  $\Delta$ , find the "best possible" gain constant  $\mu$  for the LMS that will converge within 1500 samples. Plot its convergence curve. What are the EMSE values here?

(d) Repeat part (b) for the  $\varepsilon$ -normalised LMS algorithm discussed in class. Choose an appropriate value for  $\varepsilon$  and for  $\mu$  in this case, and justify the reason for your choice(s). What are the EMSE values here?

(e) Repeat part (b), using now a Recursive Least Squares (RLS) algorithm to define the adaptive equaliser. Use forgetting factor  $\lambda$ =0.995 and initial choice of inverse of the data covariance **P**(0)=100. What is the simulated and theoretical excess MSE (EMSE) in this case?

(f) For the same  $\lambda$ =0.995, if you are allowed to change **P**(0), specify a new **P**(0) that will make RLS converge at least <u>five</u> times faster than the selection in (d). What are the EMSE values here?

(g) Make a single consolidated plot, with both the LMS curves in "solid blue",  $\varepsilon$ -normalised LMS in "solid green", and both of the RLS curves in "solid black". Mark also the J<sub>min</sub> line using "dashed red" on this same plot.