

PROBLEM 1. Consider transmission over an ISI channel where the output after matched filtering is

$$U[n] = \sum_{k=-\infty}^{\infty} d[k]I[n-k] + V[n]$$

and  $d[k]$  is given by

$$d[k] = \begin{cases} 2^{-\frac{|k|-1}{2}} & \text{if } k \text{ is odd,} \\ \frac{5}{3}2^{-\frac{|k|}{2}} & \text{if } k \text{ is even,} \end{cases}$$

and, as we already know from the lecture, the noise  $V[n]$  has autocorrelation function  $R_V[k] = \frac{N_0}{2}d[-k]$ . Find the filter  $d_W[k]$  to whiten the noise. Choose the whitening filter such that the resulting communication channel after the whitening filter is causal.

PROBLEM 2. Consider the discrete-time equivalent of a band-limited AWGN channel with noise power  $N_0/2 = 1$ . The equivalent channel is a 3-tap channel defined as

$$U[n] = \beta I[n+1] + \alpha I[n] + \beta I[n-1] + V[n].$$

The Gaussian noise  $V[n]$  has autocorrelation function

$$R_V[k] = d[-k] = \alpha\delta[k] + \beta\delta[k-1] + \beta\delta[k+1],$$

and  $\alpha$  and  $\beta$  are positive real coefficients satisfying  $\alpha^2 > 4\beta^2$ . We also assume that the information symbols  $I[n]$  are i.i.d. random variables with mean 0 and variance  $\mathcal{E}$ .

- (a) Use the zero-forcing equalizer to remove all of the inter-symbol interference. Find the frequency response of the filter and calculate the variance of the effective noise  $\tilde{V}[n] = I[n] - \hat{I}_{ZF}[n]$  (see Equation (4.13) of your lecture notes).
- (b) Now apply the LMMSE approach. Find the frequency response of LMMSE filter  $A_{LMMSE}(f)$ . Calculate the resulting noise variance  $\sigma_{LMMSE}^2$  (as in Equation (4.15) of your lecture notes) using the following steps:
  - (i) Show that, in general,  $\sigma_{LMMSE}^2 = \mathbb{E}[I[n]^2] - \mathbb{E}[\hat{I}_{LMMSE}[n]I[n]]$ .
  - (ii) Use (i) to find  $\sigma_{LMMSE}^2$  for the particular example considered in the problem.
- (c) Another tempting receiver the matched filter: It passes the channel output signal through a filter whose impulse response is the time-reversed version of the original channel impulse response. Show that the corresponding output signal can be expressed as

$$I_{MF}[n] = g[0]I[n] + \sum_{k \neq 0} g[k]I[n-k] + \sum_m f[m]V[n-m],$$

give the values of the coefficients  $g[k]$ , and  $f[m]$ , and calculate the effective noise variance, that is, the variance of the  $\frac{1}{g[0]}I_{MF}[n] - I[n]$ .

- (d) For  $\alpha = 1$  and  $\beta = 0.4$ , plot the effective noise variances you found in (a)–(c) for the range of  $\mathcal{E} \in [-20 \text{ dB}, +20 \text{ dB}]$  and compare them.

*Hint.* You may need the following integrals: For  $a \geq |b|$ ,

$$\frac{1}{2\pi} \int_{-\pi}^{\pi} \frac{dx}{a + b \cos x} = \frac{1}{\sqrt{a^2 - b^2}}, \quad \frac{1}{2\pi} \int_{-\pi}^{\pi} \frac{dx}{(a + b \cos x)^2} = \frac{a}{(a^2 - b^2)\sqrt{a^2 - b^2}}.$$

PROBLEM 3. Consider the following *two user* discrete-time channel

$$Y[n] = I[n] - 2I[n-1] + J[n] + J[n-1] + W[n]$$

where  $I[n]$  is the information symbol from user 1 and  $J[n]$  is the information symbol from user 2. We assume that the sequences  $I[n]$  and  $J[n]$  are both i.i.d. zero-mean random variables with unit variance independent of each other, and that  $W[n]$  is white Gaussian noise with variance  $N_0/2 = 1$ . Given the channel output  $Y[n]$  we would like to design linear MMSE estimators to detect both  $I[n]$  and  $J[n]$  in the form

$$\hat{I}[n] = a_{-1}Y[n+1] + a_0Y[n] + a_1Y[n-1] \quad \text{and} \quad \hat{J}[n] = b_{-1}Y[n+1] + b_0Y[n] + b_1Y[n-1].$$

Compute the optimal coefficients  $a_i, b_i$  for  $i = -1, 0, 1$  which minimize the mean squared error  $\mathbb{E}[(\hat{I}[n] - I[n])^2]$  and  $\mathbb{E}[(\hat{J}[n] - J[n])^2]$ .

PROBLEM 4. Consider the noisy ISI channel given by

$$Y[n] = X[n] + X[n-1] + Z[n]$$

where  $X[n]$  and  $Y[n]$  are the channel input and output, respectively at time index  $i$ ,  $Z[n]$  is a sequence of i.i.d. Gaussian random variables, with zero mean and unit variance and  $X[n] \in \{-1, 1\}$  uniformly. Calculate the symbol-wise MAP estimate of  $X[1], \dots, X[4]$  using the BCJR algorithm, if the received sequence is  $1, -2, -1, 3, 2$ . Assume that the channel is in state  $+1$  at the beginning and at the end of the sequence (i.e.,  $X[0] = X[5] = +1$ ). Compare this to the decoding estimate from MLSE (Viterbi) decoder.

PROBLEM 5. Consider the following real channel,

$$\mathbf{Y} = \mathbf{h}X + \mathbf{Z},$$

where  $X \in \mathbb{R}$  is a random variable, with  $\mathbb{E}[X] = 0$  and  $\mathbb{E}[X^2] = \mathcal{E}$ ,  $\mathbf{h}$  is a fixed real (column) vector, and  $\mathbf{Z}$  is a zero-mean random vector with covariance matrix  $I$ , chosen independently of  $X$ .

- (a) An estimator  $\hat{X}(\mathbf{y})$  is said to be *unbiased* if  $\mathbb{E}[\hat{X}(\mathbf{Y})|X = x] = x$ .

- (i) What is the constraint for a linear estimator, i.e  $\hat{X} = \mathbf{a}^T \mathbf{Y}$  to be unbiased?  
(ii) Find the unbiased linear estimator that minimizes the mean squared error  $\sigma_{\text{unbiased}}^2 = \mathbb{E}[(X - \hat{X})^2]$  and the value of  $\sigma_{\text{unbiased}}^2$  for this estimator.

*Hint.* By Cauchy–Schwartz inequality,  $(\mathbf{a}^T \mathbf{a})(\mathbf{h}^T \mathbf{h}) \geq |\mathbf{a}^T \mathbf{h}|^2$ .

- (b) In this part, we don't restrict ourselves to unbiased estimators. Suppose  $\hat{X} = \mathbf{a}^T \mathbf{Y}$  is a linear estimator. Find the linear estimator that minimizes the mean squared error  $\sigma^2 = \mathbb{E}[(X - \hat{X})^2]$  and the value of  $\sigma^2$  for this estimator.

*Hint.* First assume  $\mathbf{a}^T \mathbf{h} = c$ , and minimize  $\sigma^2$  with respect to the vector  $\mathbf{a}$  with the constraint  $\mathbf{a}^T \mathbf{h} = c$ , and then minimize the result with respect to  $c$ .

- (c) Compare the two “signal to noise ratio”s  $\mathcal{E}/\sigma_{\text{unbiased}}^2$  and  $\mathcal{E}/\sigma^2$ .

- (d) Now assume  $X$  is equally likely to be  $+1$  or  $-1$ . Suppose a decision is made by quantizing the estimate  $\hat{X}$  from either part (a) or (b) to  $\pm 1$ . Which estimator would you choose to minimize the probability of error?