

## ESE 471 Spring 2021: Homework 5

1. (20 points) In Lecture 10, when deriving the threshold for a 1-D binary detector, we assumed that the noise  $N$  had equal variance regardless of whether  $s_0(t)$  or  $s_1(t)$  was sent. Assume instead that the transmitter is faulty and adds noise to the transmitted signal whenever  $s_1(t)$  is sent. That is the receiver measures  $X$  as either:

$$\begin{aligned} H_0 : \quad X &= a_0 + N_0 \\ H_1 : \quad X &= a_1 + N_1 \end{aligned}$$

where  $a_0 = 0$ ,  $a_1 = 1$ , and  $N_0$  and  $N_1$  are both zero-mean Gaussian noise, in this case, where  $N_0$  has variance 1 and  $N_1$  has variance 2. Assume that  $P[H_0] = P[H_1] = 1/2$ . What is the optimal decision region  $R_0$  (the region in which we decide  $H_0$ )?

2. (10 points) Consider a binary 1-D communication system modeled as  $H_0 : X = a_0 + W$ , or  $H_1 : X = a_1 + W$ , where  $a_0 = -1$ ,  $a_1 = +1$ , and  $W$  is zero mean Gaussian with variance  $\sigma_W^2 = 0.3$ . First, assume  $P[H_0] = P[H_1] = 1/2$ .
  - (a) What is the optimal decision threshold  $\gamma$ ?
  - (b) What is the probability of error for the optimal receiver?

For the next two parts, assume that the signals are not equally probable, and instead,  $P[H_0] = 0.2$  and  $P[H_1] = 0.8$ .

- (c) What is the optimal decision threshold  $\gamma$ ?
  - (d) What is the probability of error for the optimal receiver?
3. **Optimal Detection Verification** (20 points): Continue to consider the above communication system, with  $H_0 : X = a_0 + W$ , or  $H_1 : X = a_1 + W$ , where  $a_0 = -1$ ,  $a_1 = +1$ , and  $W$  is zero mean Gaussian with variance  $\sigma_W^2 = 0.3$ , and  $P[H_0] = 0.2$  and  $P[H_1] = 0.8$ . In this problem, you will verify experimentally that the optimal decision threshold minimizes the probability of error. In Matlab or Python (or language of your choice),
  - (a) Create a vector of 5000 bits sent by the transmitter, each bit  $b_l$  is equal to 0 with probability  $P[H_0]$  and equal to 1 with probability  $P[H_1]$ . Show Matlab output of the histogram of  $b_l$  to verify its pmf.
  - (b) Generate  $X$  as follows: When  $b_l = 0$ , let  $X = a_0 + W$ . When  $b_l = 1$ , let  $X = a_1 + W$ .  $W$  is generated with  `$\sigma_W \cdot \text{randn}$`  in Matlab. In Python, use  `$\text{random.gauss}(0, \sigma_W)$`  (after having done an `import random`). No output is needed for this part.
  - (c) Create a vector listing possible threshold values, `etaList = -2:0.01:2` in Matlab or  `$\text{numpy.arange}(-2, 2.01, 0.01)$`  in Python. For each value  $\eta$  in the list, calculate the bit error rate using a threshold of  $\eta$ . Specifically, make bit decisions by the detection rule,  $X \underset{H_0}{\overset{H_1}{\gtrless}} \eta$ . Denoting these bit decisions  $\hat{b}_l$ , count the experimental bit error rate for that  $\eta$ . Plot the experimental bit error rate vs. `etaList`. Draw a vertical line on the plot at  $\eta = \gamma$ , where  $\gamma$  is the optimal threshold (see Problem 2(c)), and show that the error rate is approximately minimum at the optimal threshold.
  - (d) Do parts (a) through (c) again for the case when  $H_0$  and  $H_1$  are equally probable. The only output necessary is the final plot of experimental bit error rate vs. `etaList`.