

PROBLEM 1.

- (a)  $E[N^2] = \sum_j \alpha_j^2 \sigma_j^2$ , so we can choose  $v = (\alpha_1 \sigma_1, \dots, \alpha_n \sigma_n)$  to make  $E[N^2] = \|v\|^2$ . As  $S = \pm \sum_j \alpha_j h_j$ , choosing  $u = (h_1/\sigma_1, \dots, h_n/\sigma_n)$  will ensure  $S = \pm \langle u, v \rangle$ .
- (b) Cauchy-Schwarz inequality says  $\langle u, v \rangle^2 \leq \|u\|^2 \|v\|^2$ . As  $\|u\|^2 = \sum_j h_j^2/\sigma_j^2$ , the conclusion follows.
- (c) Equality in Cauchy-Schwarz holds if  $u$  and  $v$  are scalar multiples of each other, in particular if  $u = v$ , or equivalently, choosing  $h_j/\sigma_j = \alpha_j \sigma_j$ , again equivalently, choosing  $\alpha = (h_1/\sigma_1^2, \dots, h_n/\sigma_n^2)$ .
- (d) Note that

$$f_{Y|H}(y_1, \dots, y_n|0) = \prod_j (2\pi\sigma_j^2)^{-1/2} \exp[-(y_j - h_j)^2/(2\sigma_j^2)]$$

$$f_{Y|H}(y_1, \dots, y_n|1) = \prod_j (2\pi\sigma_j^2)^{-1/2} \exp[-(y_j + h_j)^2/(2\sigma_j^2)],$$

so the log-likelihood-ratio equals  $-2 \sum_j y_j h_j/\sigma^2$  which is  $-2t(y)$ . Since we know that the log-likelihood ratio is a sufficient statistic, we conclude that  $T$  is too.

PROBLEM 2.

- (a) Note that

$$2(2\pi\sigma^2)f_{Y|H}(y|0) = \exp(-\|y - c_0\|^2/(2\sigma^2)) + \exp(-\|y + c_0\|^2/(2\sigma^2))$$

$$2(2\pi\sigma^2)f_{Y|H}(y|1) = \exp(-\|y - c_1\|^2/(2\sigma^2)) + \exp(-\|y + c_1\|^2/(2\sigma^2)).$$

So (as  $c_0$  and  $c_1$  have the same norm) the decision rule is to decide 0 or 1 according to

$$\exp(\langle y, c_0 \rangle/\sigma^2) + \exp(-\langle y, c_0 \rangle/\sigma^2) \geq \exp(\langle y, c_1 \rangle/\sigma^2) + \exp(-\langle y, c_1 \rangle/\sigma^2).$$

As  $\langle y, c_0 \rangle = \sqrt{\mathcal{E}/2}(y_1 + y_2)$  and  $\langle y, c_1 \rangle = \sqrt{\mathcal{E}/2}(y_1 - y_2)$  the decision rule is, with  $\tilde{y}_i = \sqrt{\mathcal{E}/2}y_i/\sigma^2$ ,

$$e^{\tilde{y}_1 + \tilde{y}_2} + e^{-\tilde{y}_1 - \tilde{y}_2} - e^{\tilde{y}_1 - \tilde{y}_2} - e^{\tilde{y}_2 - \tilde{y}_1} \geq 0.$$

The left hand side above equals  $(e^{\tilde{y}_1} - e^{-\tilde{y}_1})(e^{\tilde{y}_2} - e^{-\tilde{y}_2})$ , so we decide 0 if  $y_1$  and  $y_2$  have the same sign, and decide 1 otherwise.

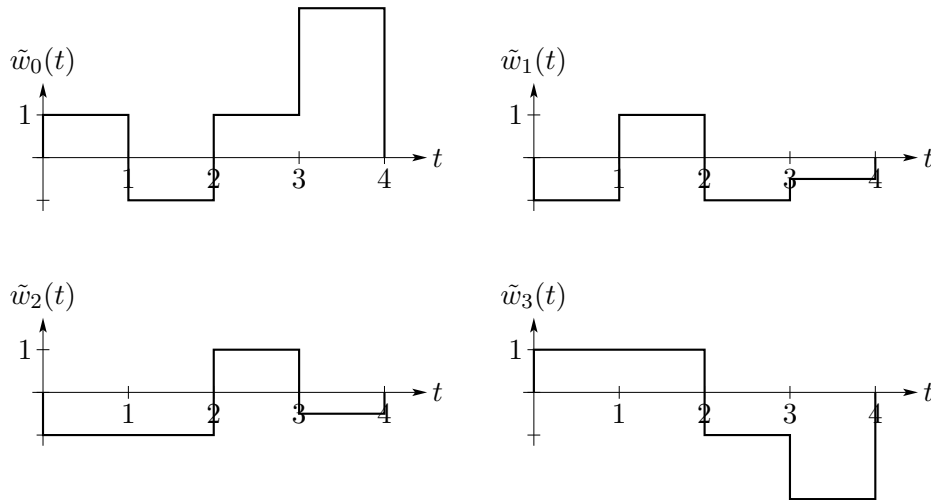
- (b) By the symmetry in the problem the error probability is the same for the two hypotheses, and is the same regardless of the value of  $A$ . So we can assume  $c_0$  is sent and  $A = 1$ . We will make an error if either  $Y_1 > 0$  and  $Y_2 < 0$ , or  $Y_1 < 0$  and  $Y_2 > 0$ . Again by symmetry, these two events have the same probability, and thus the error probability is

$$2\Pr(Z_1 > -\sqrt{\mathcal{E}/2})\Pr(Z_2 < -\sqrt{\mathcal{E}/2}) = 2Q(\sqrt{\mathcal{E}/(2\sigma^2)})[1 - Q(\sqrt{\mathcal{E}/(2\sigma^2)})].$$

- (c) If the receiver knows that  $A = 1$ , it knows that its observation is a noisy version  $c_0 = \sqrt{\mathcal{E}/2}(1, 1)$  or  $c_1 = \sqrt{\mathcal{E}/2}(1, -1)$ . The MAP rule will decide 0 if  $Y_2 > 0$  and 1 if  $Y_2 < 0$ . Similarly, if the receiver knows  $A = -1$ , it will decide 0 if  $Y_2 < 0$  and 1 if  $Y_2 > 0$ . In either case the error probability is  $Q(\sqrt{\mathcal{E}/(2\sigma^2)})$ .
- (d) If the receiver is told the correct value of  $A$ , then the error probability is the value  $q = Q(\sqrt{\mathcal{E}/(2\sigma^2)})$  we found (c). If it is told the incorrect value of  $A$  the receiver's decision regions are flipped, and thus the error probability is  $1 - q$ . Combining these, we find the error probability as  $(1 - p)q + p(1 - q)$ .

PROBLEM 3.

- (a) An orthonormal basis for the four waveforms is given by  $p(t), p(t-1), p(t-2), p(t-3)$  where  $p(t)$  is the rectangular pulse  $\mathbb{1}\{t \in [0, 1)\}$ . The map decoder would work with  $Y_1, Y_2, Y_3, Y_4$ , the inner product of  $R(t)$  with these basis functions. But these can be computed from the filter output as  $Y_1 = R_1 + R_2, Y_2 = R_3 + R_4, Y_3 = R_5 + R_6, Y_4 = R_7 + R_8$  with  $R_i$  denoting the filter output at time  $t_i = i/2$ .
- (b) The translate that gives the minimum energy should make the resulting constellation have average equal to 0. In the original constellation the average signal  $[w_0(t) + w_1(t) + w_2(t) + w_3(t)]/4$  is a piecewise constant signal taking the values 1, 2, -1, -1/2 on the intervals  $[0, 1), [1, 2), [2, 3), [3, 4)$ . The translated waveforms are obtained by subtracting this from each signal, and we obtain:



- (c) The original waveforms were meant to be such that  $w_0$  should have height -3 instead of +3. This would have made the translated waveforms form a two dimensional QAM constellation whose error probability is simple to compute. As asked, however, the translated signals have no such structure, and form a constellation in three dimensions. The error probability would be possible to compute numerically, but a closed form expression, if there is one, will be extremely ugly.
- (d) Since isometric transforms do not change the probability of error the implementation in (a) has the same (but very ugly) error probability as in (c).

PROBLEM 4.

- (a) For the case  $m = 2$ , the U–B bound is  $\iint \sqrt{f_{Y|H}(y_1, y_2|1)f_{Y|H}(y_1, y_2|2)} dy_1 dy_2$ . This evaluates as

$$\iint \sqrt{q(y_1)p(y_2)p(y_1)q(y_2)} dy_1 dy_2 = \int \sqrt{p(y_1)q(y_1)} dy_1 \int \sqrt{p(y_2)q(y_2)} dy_2 = B^2.$$

- (b) For general  $m$ , we need to first evaluate

$$\int \cdots \int \sqrt{f_{Y|H}(y_1, \dots, y_m|i)f_{Y|H}(y_1, \dots, y_m|j)} dy_1 \cdots dy_m.$$

The integrand above equals  $\sqrt{p(y_i)q(y_i)}\sqrt{p(y_j)q(y_j)}\prod_{k \neq i,j} p(y_k)$ , so the integral splits as a product of integrals. The integrals for  $y_i$  and  $y_j$  both give  $B$ , the other integrals give 1. Thus the value of the above integral is  $B^2$ . The U–B bound thus evaluates to  $(m - 1)B^2$ .

- (c) For this  $p$  and  $q$ ,  $B = \sqrt{m} \int_0^\infty \exp(-(m+1)y/2) dy = 2\sqrt{m}/(m+1)$ . The U–B bound in (b) is thus  $4m^2/(m+1)^2$ , which approaches 4 as  $m$  gets large — a rather useless bound on error probability.
- (d) The decision will be wrong only if, either  $Y_1 \leq t$ , or, for some  $j = 2, \dots, m$ ,  $Y_j > t$ . The union bound on these  $m$  events gives

$$\Pr(Y_1 \leq t) + \sum_{j=2}^m \Pr(Y_j > t)$$

as an upper bound to the probability of error. But since since  $Y_2, \dots, Y_m$  have the same distribution  $p$ ,  $\Pr(Y_j > t) = P(Y_2 > t)$ . Which yields the upper bound  $\Pr(Y_1 \leq t) + (m-1)\Pr(Y_2 > t)$ . As  $\Pr(Y_2 > t) = \exp(-t)$  and  $\Pr(Y_1 > t) = \exp(-tm)$  we obtain the desired conclusion.