

---

## MIDTERM

Wednesday, November 14, 2007, 10:15 - 13:15  
This exam has 5 problems and 100 points in total

---

- You have 3 hours.
- You are allowed to use 3 sheets of paper (6 pages) for reference.
- No other materials are allowed. No mobile phones or calculators are allowed.
- Do not spend too much time on each problem but try to collect as many points as possible.
- Write only what is relevant to the question!

**Good Luck!**

Name: \_\_\_\_\_

Prob I	/ 20
Prob II	/ 20
Prob III	/ 20
Prob IV	/ 20
Prob V	/ 20
<b>Total</b>	<b>/ 100</b>

**Problem 1** [Hypothesis Testing – 20 pts]

Consider a channel whose input  $X$  and output  $Y$  are related by

$$Y = X + Z.$$

The input  $X$  is uniformly distributed on the set  $\{-1, +1\}$ , i.e.,

$$\Pr[X = -1] = \Pr[X = +1] = 1/2.$$

The noise  $Z$  is independent from  $X$ . Given the output  $Y$ , we want to guess  $X$ .

(8 pts) (a) Suppose that  $Z$  is given by

$$Z = \begin{cases} W & \text{with probability } 1/2 \\ W + 1 & \text{with probability } 1/2 \end{cases} \quad (1)$$

where  $W$  is a random variable with density

$$f_W(w) = \begin{cases} \frac{1}{2} - \frac{|w|}{4} & \text{if } |w| \leq 2 \\ 0 & \text{otherwise} \end{cases}. \quad (2)$$

Find the decision rule that minimizes the probability of error.

(12 pts) (b) Suppose now that

$$Z = W \quad \text{or} \quad Z = W + 1, \quad (3)$$

where  $W$  is the same as in part (a), but (1) is not valid anymore. Instead, we want to find a decision rule that works well in both the cases in (3). For a given decision scheme  $H$ , define the error probability in each case by

$$\begin{aligned} P_{e,0}(H) &= \Pr[\text{error} | Z = W] \\ P_{e,1}(H) &= \Pr[\text{error} | Z = W + 1]. \end{aligned}$$

Find the decision rule that minimizes the error probability in the worst case. That is, find  $H$  that minimizes

$$\max\{P_{e,0}(H), P_{e,1}(H)\}. \quad (4)$$

**Hint 1:** Your decision rule should take the form of a threshold scheme.

**Hint 2:** What do the plots of  $P_{e,0}(t)$  and  $P_{e,1}(t)$  look like as a function of the threshold  $t$  of the decision scheme?

**Problem 2** [Proper Vectors – 20pts]

- (4 pts) (a) Consider an arbitrary real-valued signal  $x(t)$  and define  $\hat{x}(t)$  to be the signal with the Fourier transform

$$\hat{x}_{\mathcal{F}}(f) = \sqrt{2}x_{\mathcal{F}}(f)h_{>,\mathcal{F}}(f)$$

where

$$h_{>,\mathcal{F}}(f) = \begin{cases} 1 & \text{for } f > 0 \\ 1/2 & \text{for } f = 0 \\ 0 & \text{for } f < 0 \end{cases}$$

and  $x_{\mathcal{F}}(f)$  is the Fourier transform of  $x(t)$ ,  $x_{\mathcal{F}}(f) = \mathcal{F}\{x(t)\}$ .

Derive the relation that allows to go back from  $\hat{x}(t)$  to  $x(t)$ , i.e., find  $x(t)$  as a function of  $\hat{x}(t)$ .

- (4 pts) (b) Let  $\hat{Z}(f) = \sqrt{2}Z_{\mathcal{F}}(f)h_{>,\mathcal{F}}(f)$  where  $Z_{\mathcal{F}}(f) = \mathcal{F}\{Z(t)\}$  and  $Z(t)$  is a zero-mean, real-valued Gaussian and wide sense stationary process. Using the definition of a proper vector, show that  $\hat{Z}(t) = \mathcal{F}^{-1}\{\hat{Z}(f)\}$  is proper.
- (4 pts) (c) Show that  $Z_E(t) = \hat{Z}(t)e^{-j2\pi f_0 t}$  is also proper. ( $f_0$  is a positive real number.)
- (4 pts) (d) Use the previous results to show that the real and imaginary components of  $Z_E(t)$  have the same autocorrelation function.
- (4 pts) (e) Assume that  $S_Z(f_0 - f) = S_Z(f_0 + f)$  where  $S_Z(f)$  is the Fourier transform of the autocorrelation function of  $Z_E(t)$ . Show that the real and imaginary parts of  $Z_E(t)$  are independent.

**Problem 3** [Viterbi Decoder – 20pts]

Let  $\{U_1, \dots, U_n\}$  be a sequence of i.i.d. random variables over  $\{0, 1\}$  with probability distribution  $P(0) = P(1) = 1/2$ . Consider the following channel. The input and output of the channel are binary  $\{0, 1\}$  with the following transition probabilities

$$\begin{aligned} P_{Y|X}(1 | 1) &= 1 - \epsilon, \\ P_{Y|X}(0 | 1) &= \epsilon, \\ P_{Y|X}(0 | 0) &= 1 - \epsilon, \\ P_{Y|X}(1 | 0) &= \epsilon. \end{aligned}$$

Let  $f(x, y)$  denote the transition probability  $P_{Y|X}(y | x)$ .

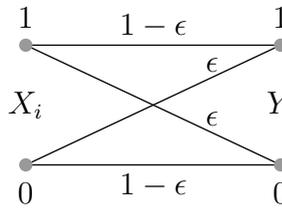


Figure 1: The graph showing transmission between source and receiver

Let  $U_0 = 0$ . For each bit  $U_i$ ,  $1 \leq i \leq n$  we transmit two bits  $X_{2i-1} = U_i$  and  $X_{2i} = U_i + U_{i-1}$ , where the addition here means XOR (modulo 2 addition). Let the source vector be  $\{U_1, \dots, U_n\}$ , the transmitted vector be  $\{X_1, \dots, X_{2n}\}$  and the received vector be  $\{Y_1, \dots, Y_{2n}\}$ .

- (5 pts) (a) Write the MAP rule for decoding the sequence  $\{U_1, \dots, U_n\}$  from  $\{Y_1, \dots, Y_{2n}\}$  in terms of the function  $f$ .
- (10 pts) (b) Construct a Viterbi decoder for this MAP rule. Use log domain calculation. Let  $y_{2i-1} = 0, y_{2i} = 1$ . Draw the trellis section related to the bit  $U_i$  and compute the branch metrics in terms of  $\epsilon$ .
- (5 pts) (c) Consider the case where the priors are not uniform. Let  $P(0) = p_0$ ,  $P(1) = p_1 = 1 - p_0$ . What are the new branch metrics?

**Problem 4** [Estimation – 20pts]

Suppose we observe the sequence  $\{y(k)\}$  and we want to estimate  $\{x(k)\}$ . For this purpose, we use a linear filter  $w(k)$ . Let  $\hat{x}(k)$  be the estimate of  $x(k)$ :

$$\hat{x}(k) = w(k) * y(k).$$

Let  $w_{opt}(k)$  be the optimal estimator according to the MMSE criterion, i.e. the estimator minimizing  $\mathbb{E}[|e(k)|^2]$ , where  $e(k) = x(k) - \hat{x}(k)$ .

Show that the optimal estimator  $w_{opt}(k)$  is minimizing  $\mathbb{E}[|e(k)|^2]$  *if and only if* the following condition is met:

$$\mathbb{E}[e_{opt}(k)y^*(k-n)] = 0, \text{ for all } n$$

where  $e_{opt}(k) = x(k) - \hat{x}_{opt}(k)$  and  $\hat{x}_{opt}(k) = w_{opt}(k) * y(k)$ .

Provide a complete proof, i.e., prove both “if” and “only if” parts.

Hint: start by evaluating the performance on an estimator which does not satisfy the condition above.

**Problem 5** [Equalization – 20pts]

Consider the channel:

$$y(k) = |p|x(k) * q(k) + z(k),$$

where the input symbols  $\{x(k)\}$  are i.i.d. and independent of the noise samples  $\{z(k)\}$ . We denote the energy of the transmitted symbols by  $\mathcal{E}_x = \mathbb{E}[|x(k)|^2]$ . The noise samples  $\{z(k)\}$  have the autocorrelation function  $R_z(k) = q(k)N_0$ .

The channel coefficients have the property that  $(q(n))^* = q(-n)$ , and  $p$  is a constant scalar.

- (12 pts) (a) Using time domain calculations, compute  $r_{xy}(n) = \mathbb{E}[x(k)y^*(k-n)]$  and  $r_{yy}(n) = \mathbb{E}[y(k)y^*(k-n)]$  in terms of  $q(n)$ .
- (3 pts) (b) Compute the cross spectrum and power spectrum  $S_{xy}(D) = \mathcal{D}\{r_{xy}(n)\}$  and  $S_{yy}(D) = \mathcal{D}\{r_{yy}(n)\}$  respectively.
- (5 pts) (c) We convolve  $y(k)$  with a filter  $w(k)$  in order to estimate  $x(k)$  and define the estimation error  $e(k) = x(k) - w(k) * y(k)$ . In order to minimize this error according to the MMSE criterion, we need to satisfy  $\mathbb{E}[e_{opt}(k)y^*(k-n)] = 0$  for all  $n$ , where  $e_{opt}(k) = x(k) - w_{opt}(k) * y(k)$ . Use this condition to derive the optimal filter  $w_{opt}(k)$ . Using the previous parts of the exercise compute the optimal filter in D-transform domain.