

## EEE 431: Telecommunications 1

### Quiz 3

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Prob. 1: \_\_\_\_\_ / 40

Prob. 2: \_\_\_\_\_ / 35

Prob. 3: \_\_\_\_\_ / 25

**Total: \_\_\_\_\_ / 100**

Some trigonometric identities:  $\sin(2x) = 2 \sin(x) \cos(x)$

$$\cos(2x) = 1 - 2 \sin^2(x) = 2 \cos^2(x) - 1$$

$$\cos(x + y) = \cos(x) \cos(y) - \sin(x) \sin(y)$$

$$\sin(x + y) = \sin(x) \cos(y) + \cos(x) \sin(y)$$

$$\sin(x) \cos(y) = 0.5 \sin(x + y) + 0.5 \sin(x - y)$$

$$\cos(x) \cos(y) = 0.5 \cos(x + y) + 0.5 \cos(x - y)$$

$$\sin(x) \sin(y) = 0.5 \cos(x - y) - 0.5 \cos(x + y)$$

$$\sin(\pi/6) = \cos(\pi/3) = 1/2, \quad \sin(\pi/3) = \cos(\pi/6) = \sqrt{3}/2.$$

**Problem 1** Consider the following random process:  $X(t) = A \cos(2000\pi \alpha t + \theta)$ , where  $A > 0$  is a known fixed number,  $\theta$  is a uniform random variable in the interval of  $[0, 2\pi)$ , and  $\alpha$  is a discrete random variable, which is equal to 1 or 2 with equal probabilities.

- (a) Find the autocorrelation function of  $X(t)$ .
- (b) Find the power spectral density of  $X(t)$ .
- (c) Calculate the average power of  $X(t)$ .
- (d) Suppose that  $X(t)$  passes through a linear time-invariant filter with the following frequency response:

$$H(f) = \begin{cases} (3000 - |f|)/1000, & \text{if } |f| \leq 3000 \\ 0, & \text{otherwise} \end{cases}$$

Find the autocorrelation function of the output of the filter.

Hint: The Fourier transform  $S(f)$  of  $s(t)$  is defined as  $S(f) = \int_{-\infty}^{\infty} s(t)e^{-j2\pi ft}dt$  and the inverse Fourier transform is given by  $s(t) = \int_{-\infty}^{\infty} S(f)e^{j2\pi ft}df$ .

**Problem 2** Consider a binary communication system operating over an additive white Gaussian noise channel with spectral density level of  $N_0/2$ . The transmitted signals are given by

$$s_1(t) = \sqrt{\frac{8}{T}} \cos\left(2\pi f_c t - \frac{\pi}{6}\right), \quad s_2(t) = \sqrt{\frac{2}{T}} \sin\left(2\pi f_c t + \frac{4\pi}{3}\right)$$

for  $t \in [0, T]$ , and they are zero otherwise. Suppose that  $f_c T$  is an integer.

- (a) Calculate the energy of  $s_1(t)$  and the energy of  $s_2(t)$ .
- (b) Find orthonormal basis function(s) to represent these signals, and illustrate them as constellation points in the corresponding signal space.
- (c) Show the structure of the optimal receiver.
- (d) Calculate the probability of error of the optimal receiver.

**Problem 3** Consider a binary communication system, where the vector representations of the transmitted signals are denoted by  $\mathbf{s}_1$  and  $\mathbf{s}_2$ . For simplicity, you can assume that each vector is 2-dimensional. The system operates over an additive white Gaussian noise channel with spectral density level  $N_0/2$ . That is, the observation vector at the receiver is  $\mathbf{r} = \mathbf{s}_i + \mathbf{n}$  for  $i \in \{1, 2\}$ , where the components of  $\mathbf{n}$  are independent and identically distributed zero mean Gaussian random variables each with variance  $N_0/2$ . At the receiver, the maximum likelihood (ML) decision rule processes  $\mathbf{r}$  and makes a decision about the transmitted signal. Let  $P_e$  denote the probability of error for this receiver. Now suppose that instead of  $\mathbf{s}_1$  and  $\mathbf{s}_2$ , we use the following signals:

$$\tilde{\mathbf{s}}_1 = k\mathbf{s}_1 + \mathbf{a}, \quad \tilde{\mathbf{s}}_2 = k\mathbf{s}_2 + \mathbf{a}$$

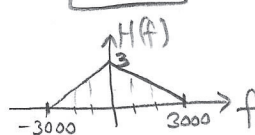
where  $k > 0$  is a known scalar, and  $\mathbf{a}$  is a known 2-dimensional vector. Considering the use of the ML decision rule based on  $\tilde{\mathbf{r}} = \tilde{\mathbf{s}}_i + \mathbf{n}$ , express the probability of error at the receiver in terms of  $P_e$  and any other parameter(s). (The probability that a zero-mean, unit variance Gaussian random variable is larger than  $x$  is defined as  $Q(x)$ , i.e.,  $Q(x) = \int_x^{\infty} \frac{1}{\sqrt{2\pi}} e^{-t^2/2} dt$ . The  $Q$ -function is monotone decreasing.)

①  $X(t) = A \cos(2000\pi t + \alpha + \theta)$   $\theta \sim \mathcal{U}[0, 2\pi)$   $p(\alpha) = \begin{cases} 0.5, & \text{if } \alpha = 1 \\ 0.5, & \text{if } \alpha = 2 \end{cases}$

13 a)  $R_X(t_1, t_2) = E[X(t_1)X(t_2)] = \frac{A^2}{2} E[\cos(2000\pi\alpha(t_1+t_2) + 2\theta) + \cos(2000\pi\alpha(t_1-t_2))]$   
 $\xrightarrow{t=t_1-t_2} = \frac{A^2}{2} \left( \frac{1}{2} \cos(2000\pi(t_1-t_2)) + \frac{1}{2} \cos(4000\pi(t_1-t_2)) \right)$

b)  $S_X(f) = F\{R_X(\tau)\} = \frac{A^2}{8} (\delta(f-1000) + \delta(f+1000) + \delta(f-2000) + \delta(f+2000))$

c)  $P_X = R_X(0) = \int_{-\infty}^{\infty} S_X(f) df = \frac{A^2}{2}$

d)  $X(t) \rightarrow \boxed{H(f)} \rightarrow Y(t)$   
  
 $S_Y(f) = S_X(f) |H(f)|^2$   
 $= \frac{A^2}{2} (\delta(f-1000) + \delta(f+1000)) + \frac{A^2}{8} (\delta(f-2000) + \delta(f+2000))$   
 $R_Y(\tau) = \frac{A^2}{4} \cos(2000\pi\tau) + \frac{A^2}{4} \cos(4000\pi\tau)$

② a)  $E_1 = \int_0^T \frac{8}{T} \cos^2(2\pi f_c t - \frac{\pi}{6}) dt = \frac{4}{T} \int_0^T (1 + \cos(4\pi f_c t - \frac{\pi}{3})) dt = 4$  Similarly,  $E_2 = 1$ .

b)  $s_1(t) = 2\sqrt{\frac{2}{T}} \left( \frac{\sqrt{3}}{2} \cos(2\pi f_c t) + \frac{1}{2} \sin(2\pi f_c t) \right)$ ,  $s_2(t) = \sqrt{\frac{2}{T}} \left( -\frac{1}{2} \sin(2\pi f_c t) - \frac{\sqrt{3}}{2} \cos(2\pi f_c t) \right)$


So,  $s_1(t) = -2s_2(t)$ . Choose basis as  $\{s_1(t), s_2(t)\}$  has unit energy.

Then,  $s_1 = [-2]$ ,  $s_2 = [1]$

  $\leftarrow$  binary PAM

c)  $r(t) \rightarrow \otimes \xrightarrow{\gamma(t)} \int_0^T \rightarrow r_1 \rightarrow \boxed{\text{ML rule}} \rightarrow \hat{m} = \begin{cases} 1, & \text{if } r_1 < -0.5 \\ 2, & \text{if } r_1 > -0.5 \end{cases}$

d)  $P_e = P_{e,1} = P_{e,2} = P(r_1 > -0.5 | s_1(t) \text{ sent}) = P(-2 + n_1 > -0.5) = Q\left(\frac{1.5}{\sqrt{N_0/2}}\right)$

③  $r = s_i + n$   $i \in \{1, 2\} \rightarrow P_e = Q\left(\frac{\|s_1 - s_2\|/2}{\sqrt{N_0/2}}\right) = Q\left(\frac{\|s_1 - s_2\|}{\sqrt{2N_0}}\right)$  (\*) 

$\tilde{r} = \tilde{s}_i + n$   $i \in \{1, 2\} \rightarrow \tilde{P}_e = Q\left(\frac{\|\tilde{s}_1 - \tilde{s}_2\|}{\sqrt{2N_0}}\right) = Q\left(\frac{k \|s_1 - s_2\|}{\sqrt{2N_0}}\right)$

From (\*),  $Q^{-1}(P_e) = \frac{\|s_1 - s_2\|}{\sqrt{2N_0}}$ . So,  $\tilde{P}_e = Q\left(k Q^{-1}(P_e)\right)$ .