

## EEE 431: Telecommunications 1

### Quiz 3

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Section: \_\_\_\_\_

Prob. 1: \_\_\_\_\_ / 30

Prob. 2: \_\_\_\_\_ / 30

Prob. 3: \_\_\_\_\_ / 40

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

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

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

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

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

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

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

**Problem 1** This question has two independent parts:

(a) Consider a signal  $s(t)$  that is defined over a duration of  $[0, T]$  seconds. Suppose that  $\psi_1(t)$  and  $\psi_2(t)$  are two orthonormal basis functions for  $s(t)$ ; i.e.,  $s(t)$  resides in the signal space generated by  $\psi_1(t)$  and  $\psi_2(t)$ . Let the vector representation of  $s(t)$  in this signal space be denoted by  $\mathbf{s}$ . Prove or disprove the following statement: “The energy of  $s(t)$  is always equal to the inner product of  $\mathbf{s}$  with itself; i.e.,  $\mathbf{s}^T \mathbf{s}$ .”

(b) Write down expressions for three time domain signals  $s_1(t)$ ,  $s_2(t)$ , and  $s_3(t)$ , defined over a duration of  $[0, T]$  seconds, which satisfy all of the following conditions:

(i) The carrier frequencies of the signals are all equal to  $f_c$  Hz, where  $f_c T$  is an integer.

(ii)  $s_2(t)$  is orthogonal to both  $s_1(t)$  and  $s_3(t)$ .

(iii)  $s_1(t)$  and  $s_3(t)$  have the same energy.

(iv) The distance between  $s_1(t)$  and  $s_2(t)$  is equal to the distance between  $s_1(t)$  and  $s_3(t)$ .

(Your answer may not be unique.)

**Problem 2** This question has two independent parts:

(a) Consider a ternary communications system with the following signals:

$$\mathbf{s}_1 = \begin{bmatrix} \sqrt{2} \\ 0 \\ 0 \end{bmatrix}, \quad \mathbf{s}_2 = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \quad \mathbf{s}_3 = \begin{bmatrix} 0 \\ 0 \\ 2 \end{bmatrix}$$

Write down a mathematical expression for the maximum likelihood (ML) decision rule at the optimal receiver, and simplify it as much as possible. (Let  $r_i$  denote the  $i$ -th correlator output at the optimal receiver.)

(b) Consider a binary communications system with the following signals:

$$\mathbf{s}_1 = \begin{bmatrix} 1 \\ 2 \end{bmatrix}, \quad \mathbf{s}_2 = \begin{bmatrix} 5 \\ 5 \end{bmatrix}$$

Considering the ML decision rule, calculate the conditional probability of error for message 1. In other words, find the probability of error when  $\mathbf{s}_1$  is the transmitted signal. The noise at the correlator outputs is modeled as independent and identically distributed zero mean Gaussian random variables with variance  $N_0/2$ . (You do not have to write down the ML decision rule.) (The probability that a zero-mean, unit variance Gaussian random variable is larger than  $x$  is defined as  $Q(x)$ , i.e.,  $Q(x) = \int_x^\infty \frac{1}{\sqrt{2\pi}} e^{-t^2/2} dt$ .)

**Problem 3** Consider two real-valued, independent, and wide-sense stationary (WSS) random processes  $W(t)$  and  $N(t)$ , where  $W(t)$  is a zero-mean white Gaussian process with a power spectral density level of  $\eta$  for all frequencies, and  $N(t)$  is a Gaussian process with mean  $\mu$  and autocorrelation function  $R_N(\tau) = \sigma \text{sinc}(\tau)$ . Define a new random process as  $X(t) = \alpha W(t) + \beta N(t)$ .

(a) Find the autocorrelation function of  $X(t)$ .

(b) Find the power spectral density of  $X(t)$ , and plot it (mark all the values on the plot).

(c) Calculate the average power of  $X(t)$ .

(d) Suppose that  $X(t)$  passes through an ideal low-pass filter with a cut-off frequency of 10 Hz, and let  $Y(t)$  denote the output of this filter. Specify the probability density function of  $Y(5)$ .

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  $s(t) = \int_{-\infty}^\infty S(f)e^{j2\pi ft} df$ .

① a)  $\underline{s} = \begin{bmatrix} s_1 \\ s_2 \end{bmatrix}$   $s_1 = \int_0^T s(t) \gamma_1(t) dt$   $s_2 = \int_0^T s(t) \gamma_2(t) dt$   $s(t) = s_1 \gamma_1(t) + s_2 \gamma_2(t)$

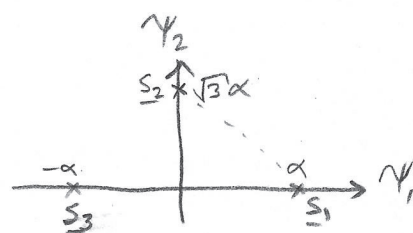
$$E_s = \int_0^T s^2(t) dt = \int_0^T (s_1 \gamma_1(t) + s_2 \gamma_2(t))^2 dt = s_1^2 \underbrace{\int_0^T \gamma_1^2(t) dt}_1 + 2 s_1 s_2 \underbrace{\int_0^T \gamma_1(t) \gamma_2(t) dt}_0 + s_2^2 \underbrace{\int_0^T \gamma_2^2(t) dt}_1$$

$$= s_1^2 + s_2^2 = \underline{s}^T \underline{s} \quad \checkmark$$

② b)  $s_1(t) = A \cos(2\pi f_c t)$   
 $s_2(t) = \sqrt{3} A \sin(2\pi f_c t)$   $t \in [0, T]$   
 $s_3(t) = -A \cos(2\pi f_c t)$

$$\|\underline{s}_1 - \underline{s}_2\| = \|\underline{s}_1 - \underline{s}_3\|$$

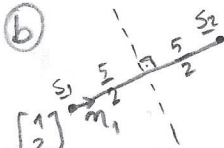
$2\alpha \qquad 2\alpha$



② a) ML rule:  $\hat{m} = \arg \min_{i \in \{1, 2, 3\}} \|\underline{r} - \underline{s}_i\|^2 = \arg \max_{i \in \{1, 2, 3\}} \left( \underline{r}^T \underline{s}_i - \frac{\|\underline{s}_i\|^2}{2} \right)$

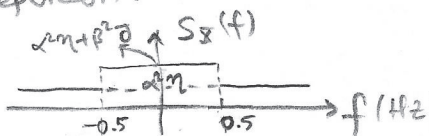
$$\hat{m} = \begin{cases} 1, & \sqrt{2}r_1 - 1 \geq r_2 - \frac{1}{2} \text{ \& } \sqrt{2}r_1 - 1 \geq r_3 - 2 \\ 2, & r_2 - \frac{1}{2} > \sqrt{2}r_1 - 1 \text{ \& } r_2 - \frac{1}{2} \geq r_3 - 2 \\ 3, & r_3 - 2 > \sqrt{2}r_1 - 1 \text{ \& } r_3 - 2 > r_2 - \frac{1}{2} \end{cases}$$

$\sqrt{2}r_1 - 1, i=1$   
 $r_2 - \frac{1}{2}, i=2$   
 $r_3 - 2, i=3$

② b)   $P_{e,1} = P_{e,2} = P_e = P\left(n_1 > \frac{5}{2}\right) = Q\left(\frac{5/2}{\sqrt{N_0/2}}\right)$

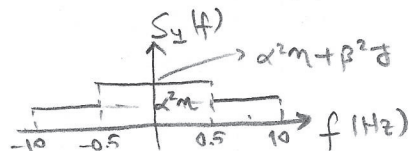
③  $\underline{y}(t) = \alpha \underline{w}(t) + \beta \underline{n}(t)$

①  $R_{\underline{y}}(\tau) = E[(\alpha \underline{w}(t+\tau) + \beta \underline{n}(t+\tau))(\alpha \underline{w}(t) + \beta \underline{n}(t))]$   
 $= \alpha^2 R_w(\tau) + \alpha \beta E[\underline{w}(t+\tau) \underline{n}(t)] + \alpha \beta E[\underline{n}(t+\tau) \underline{w}(t)] + \beta^2 R_n(\tau)$   
 $= \boxed{\alpha^2 \eta \delta(\tau) + \beta^2 \sigma \sin(\tau)}$   $\hookrightarrow$  zero since  $\underline{w}(t)$  is zero mean &  $\underline{w}(t)$  &  $\underline{n}(t)$  are independent.

②  $S_{\underline{y}}(f) = \mathcal{F}\{R_{\underline{y}}(\tau)\} = \boxed{\alpha^2 \eta + \beta^2 \sigma \Pi(f)}$  

③  $P_{\underline{y}} = \int_{-\infty}^{\infty} S_{\underline{y}}(f) df = \boxed{\infty}$

④  $S_{\underline{y}}(f) = |H(f)|^2 S_{\underline{y}}(f)$



$$E[\underline{y}(t)] = H(0) E[\underline{z}(t)] = 1 \left( \alpha \underbrace{E[\underline{w}(t)]}_0 + \beta \underbrace{E[\underline{n}(t)]}_\mu \right) = \underline{\beta \mu}$$

$$E[\underline{y}^2(t)] = R_{\underline{y}}(0) = \int_{-\infty}^{\infty} S_{\underline{y}}(f) df = (\alpha^2 \eta + \beta^2 \sigma) + 19 \alpha^2 \eta = \underline{20 \alpha^2 \eta + \beta^2 \sigma}$$

$$\underline{y}(t) \sim N(\underline{\beta \mu}, 20 \alpha^2 \eta + \beta^2 \sigma - \beta^2 \mu^2)$$

$\downarrow$   
Gaussian