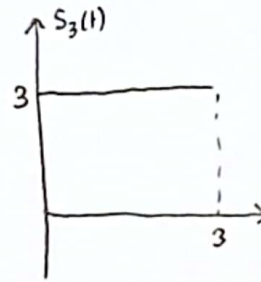
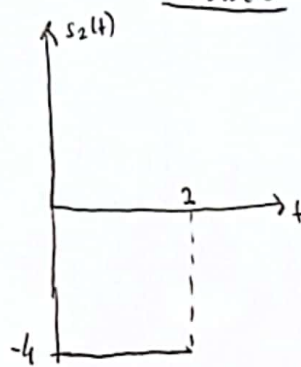
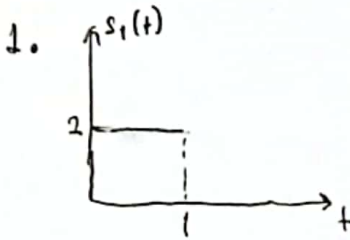


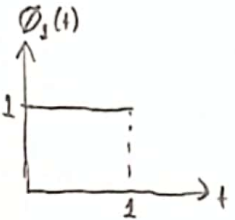
EE451 Recitation - 1

Answers



a) Find set of orthonormal basis function

$$E_1 = \int_{-\infty}^{\infty} (s_1(t))^2 dt = 4 \quad \phi_1(t) = \frac{s_1(t)}{\sqrt{E_1}} = \begin{cases} 1, & 0 \leq t \leq 1 \\ 0, & \text{o.w} \end{cases}$$



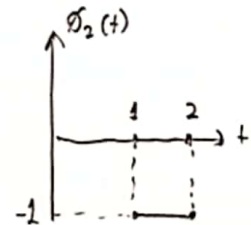
c_{21} : projection of $s_2(t)$ on $\phi_1(t)$

$$c_{21} = \langle s_2(t), \phi_1(t) \rangle = \int_{-\infty}^{\infty} s_2(t) \phi_1^*(t) dt = -4$$

$y_2(t)$ \rightarrow orthogonal part of $s_2(t)$ (but not orthonormal)

$$y_2(t) = s_2(t) - c_{21} \phi_1(t) = \begin{cases} -4 & 1 \leq t \leq 2 \\ 0 & \text{o.w} \end{cases}$$

$$E_2 = \int_{-\infty}^{\infty} (y_2(t))^2 dt = 16 \quad \phi_2(t) = \frac{y_2(t)}{\sqrt{E_2}} = \begin{cases} -1 & 1 \leq t \leq 2 \\ 0 & \text{o.w} \end{cases}$$

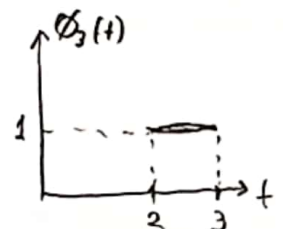


$$c_{31} = \langle s_3(t), \phi_1(t) \rangle = \int_{-\infty}^{\infty} s_3(t) \phi_1(t) dt = 0$$

$$c_{32} = \langle s_3(t), \phi_2(t) \rangle = \int_{-\infty}^{\infty} s_3(t) \phi_2(t) dt = -3$$

$$y_3(t) = s_3(t) - c_{31} \phi_1(t) - c_{32} \phi_2(t) = \begin{cases} 3, & 2 \leq t \leq 3 \\ 0, & \text{o.w} \end{cases}$$

$$E_3 = \int_{-\infty}^{\infty} (y_3(t))^2 dt = 9 \quad \phi_3(t) = \frac{y_3(t)}{\sqrt{E_3}} = \begin{cases} 1, & 2 \leq t \leq 3 \\ 0, & \text{o.w} \end{cases}$$



$$s_1 = [\sqrt{E_1} \ 0 \ 0] = [2 \ 0 \ 0] \quad s_1(t) = 2 \phi_1(t)$$

$$s_2 = [c_{21} \ \sqrt{E_2} \ 0] = [-4 \ 4 \ 0] \quad s_2(t) = -4 \phi_1(t) + 4 \phi_2(t)$$

$$s_3 = [c_{31} \ c_{32} \ \sqrt{E_3}] = [0 \ -3 \ 3] \quad s_3(t) = 0 \phi_1(t) - 3 \phi_2(t) + 3 \phi_3(t)$$

2. You can choose starting signal at any order, but selecting the shortest duration signal generally makes the computation easier.

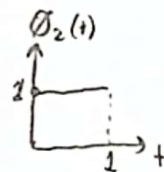
In the lecture notation confused some of the students so I will give the orthonormal set ids same as the matching signal ids (for $s_2(t)$ it will be $\phi_2(t)$)

We can start from $s_2(t)$:

$$\phi_2(t) = \frac{s_2(t)}{\sqrt{E_2}}$$

$$E_2 = \int_{-\infty}^{\infty} s_2^2(t) dt = 4$$

$$\phi_2(t) = \begin{cases} 1 & 0 \leq t \leq 1 \\ 0 & \text{o.w} \end{cases}$$



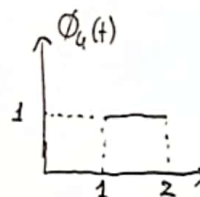
we can select the second signal as $s_4(t)$ for easier computation:

$$c_{42} : \text{projection of } s_4(t) \text{ on to } \phi_2(t) \quad \int_{-\infty}^{\infty} s_4(t) \phi_2^*(t) dt = 2$$

$$y_4(t) = s_4(t) - c_{42} \phi_2(t) = \begin{cases} 2, & 1 \leq t \leq 2 \\ 0, & \text{o.w} \end{cases}$$

$$E_4 = \int_{-\infty}^{\infty} (s_4(t))^2 dt = 4$$

$$\phi_4(t) = \begin{cases} 1, & 1 \leq t \leq 2 \\ 0, & \text{o.w} \end{cases}$$



Select $s_1(t)$

$$c_{12} : \text{projection of } s_1(t) \text{ on to } \phi_2(t) = \int_{-\infty}^{\infty} s_1(t) \phi_2^*(t) dt = 2$$

$$c_{14} : \text{ " " " " } \phi_4(t) = \int_{-\infty}^{\infty} s_1(t) \phi_4^*(t) dt = 2$$

$$y_1(t) = s_1(t) - c_{12} \phi_2(t) - c_{14} \phi_4(t) = \begin{cases} 2, & 2 \leq t \leq 3 \\ 0, & \text{o.w} \end{cases}$$

$$E_1 = \int_{-\infty}^{\infty} (s_1(t))^2 dt = 4$$

$$\phi_1(t) = \frac{y_1(t)}{\sqrt{E_1}} = \begin{cases} 1, & 2 \leq t \leq 3 \\ 0, & \text{o.w} \end{cases}$$

$s_3(t)$ can fully be represented by found orthonormal set

Now you may rename

$$\phi_2(t) \rightarrow \phi_1(t)$$

$$\phi_4(t) \rightarrow \phi_2(t)$$

$$\phi_1(t) \rightarrow \phi_3(t) \quad \text{to avoid confusion}$$

$$s_1 = [2 \quad 2 \quad 2]$$

$$s_2 = [2 \quad 0 \quad 0]$$

$$s_3 = [0 \quad -2 \quad -2]$$

$$s_4 = [2 \quad 2 \quad 0]$$

distance between any vector:

$$d_{mn} = \sqrt{\|s_m - s_n\|^2}$$

$$d_{\min} = \min_{m \neq n} (d_{mn}) = 2$$

$$d_{12} = \sqrt{0}$$

$$d_{13} = 6$$

$$d_{14} = 2$$

$$d_{23} = 2\sqrt{3}$$

$$d_{24} = 2$$

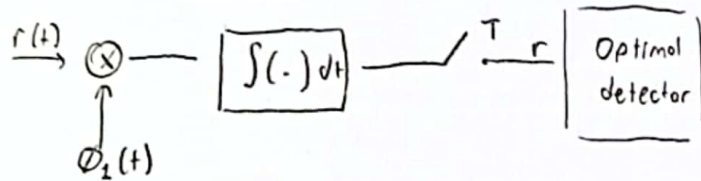
$$d_{34} = 2\sqrt{6}$$

$$3. \phi_1(t) = \frac{s_1(t)}{\sqrt{E_1}} \quad E_1 = A^2 T \quad \phi_1(t) = \begin{cases} \frac{1}{\sqrt{T}} & 0 \leq t \leq T \\ 0 & \text{o.w} \end{cases}$$

since we have only one dimension

$$s_0 = 0$$

$$s_1 = A\sqrt{T}$$



$$r = \int_0^T r(t) \phi_1(t) dt \quad \text{where } r(t) = s_m(t) + n(t)$$

$$= \int_0^T s_m(t) \phi_1(t) dt + \int_0^T n(t) \phi_1(t) dt$$

$$= s_m + \frac{1}{\sqrt{T}} \int_0^T n(t) dt$$

has zero mean and variance of :

$$\sigma_n^2 = E \left[\frac{1}{\sqrt{T}} \int_0^T n(t) dt + \frac{1}{\sqrt{T}} \int_0^T n(z) dz \right]$$

$$\sigma_n^2 = \frac{1}{T} \int_0^T \int_0^T E[n(t)n(z)] dt dz$$

$$= \frac{1}{T} \int_0^T \int_0^T \frac{N_0}{2} \delta(t-z) dt dz$$

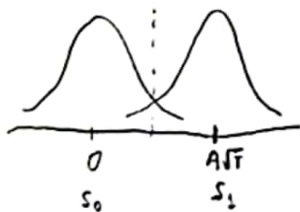
$$= \frac{1}{T} \int_0^T \frac{N_0}{2} dt = \frac{N_0}{2}$$

AWGN variance cheat sheet yaz

Gaussian pdf with mean s_m and $\sigma = \sqrt{\frac{N_0}{2}}$

$$f(r|s_m) = \frac{1}{\sqrt{\pi N_0}} e^{-\frac{(r-s_m)^2}{N_0}}$$

Since prior probabilities are equal



$$f(r|s_0) \stackrel{s_0}{\geq} f(r|s_1)$$

$$\frac{1}{\sqrt{\pi N_0}} e^{-\frac{r^2}{N_0}} \stackrel{s_0}{\geq} \frac{1}{\sqrt{\pi N_0}} e^{-\frac{(r-A\sqrt{T})^2}{N_0}} \Rightarrow e^{\frac{(r-A\sqrt{T})^2 - r^2}{N_0}} \stackrel{s_0}{\geq} 1$$

$$r \stackrel{s_1}{\geq} \frac{1}{2} A\sqrt{T}$$

optimum threshold

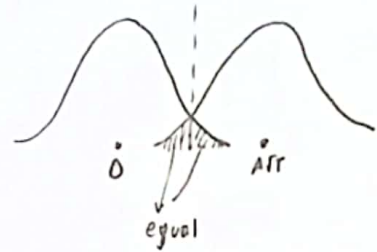
optimum threshold is $\frac{1}{2} A\sqrt{T}$

3.b.

probability of error

$$P_e = P(e|s_0)P(s_0) + P(e|s_1)P(s_1)$$

$$= \frac{1}{2} \int_{\frac{1}{2}A\sqrt{T}}^{\infty} f(r|s_0) dr + \frac{1}{2} \int_{-\infty}^{\frac{1}{2}A\sqrt{T}} f(r|s_1) dr$$



→ Due to symmetry we can compute one of them and multiply by 2

$$= 2 \cdot \frac{1}{2} \int_{\frac{1}{2}A\sqrt{T}}^{\infty} f(r|s_0) dr = \int_{\frac{1}{2}A\sqrt{T}}^{\infty} \frac{1}{\sqrt{\pi N_0}} e^{-\frac{r^2}{N_0}} dr$$

→ convert to standard normal
 $z = \frac{r - \mu}{\sigma}$

$$r = \sqrt{\frac{N_0}{2}} z$$

$$dr = \sqrt{\frac{N_0}{2}} dz$$

$$r = \frac{1}{2}A\sqrt{T} \Rightarrow \sqrt{\frac{N_0}{2}} z = \frac{1}{2}A\sqrt{T}$$

$$= \int_{\frac{A\sqrt{T}}{\sqrt{2N_0}}}^{\infty} \frac{1}{\sqrt{\pi N_0}} e^{-\frac{N_0}{2} z^2} \sqrt{\frac{N_0}{2}} dz$$

$$P_e = \int_{\frac{A\sqrt{T}}{\sqrt{2N_0}}}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}} dz = Q\left[\frac{A\sqrt{T}}{\sqrt{2N_0}}\right]$$

$$E_b = E_{avg} = \frac{A^2 T}{2} \Rightarrow$$

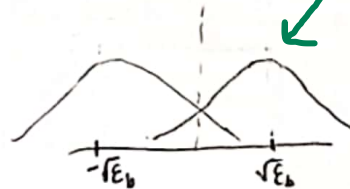
$$P_e = Q\left[\sqrt{\frac{E_b}{N_0}}\right]$$

antipodal demek birisi A ise birisi -A o yüzden bu signal constellation grafii çkyor

4. Maximum likelihood decision rule selects m that maximizes $f(r|s_m)$

$$\text{so } f(r|s_0) \stackrel{s_0}{\underset{s_1}{>}} f(r|s_1)$$

$$e^{-\frac{(r+\sqrt{E_b})^2}{N_0}} \stackrel{s_0}{\underset{s_1}{>}} e^{-\frac{(r-\sqrt{E_b})^2}{N_0}}$$



$$E_{b,avg} = \frac{E_{avg}}{k} = A^2 T$$

$$e^{\frac{(r-\sqrt{E_b})^2 - (r+\sqrt{E_b})^2}{N_0}} \stackrel{s_0}{\underset{s_1}{>}} 1$$

$$\frac{-4r\sqrt{E_b}}{N_0} \stackrel{s_0}{\underset{s_1}{>}} 0 \quad r \stackrel{s_1}{\underset{s_0}{>}} 0$$

yanlıs

$$b) P(e) = P(e|s_1)P(s_1) + P(e|s_0)P(s_0)$$

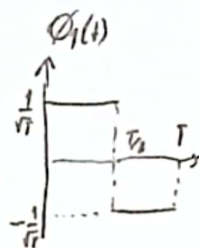
$$P_e = p \int_{-\infty}^0 \frac{1}{\sqrt{\pi N_0}} e^{-\frac{(r-\sqrt{E_b})^2}{N_0}} dr + (1-p) \int_0^{\infty} \frac{1}{\sqrt{\pi N_0}} e^{-\frac{(r+\sqrt{E_b})^2}{N_0}} dr$$

$$P_e = \frac{p}{\sqrt{2\pi}} \int_{-\frac{\sqrt{E_b}}{\sqrt{N_0/2}}}^0 e^{-\frac{z^2}{2}} dz + (1-p) \int_0^{\frac{\sqrt{E_b}}{\sqrt{N_0/2}}} \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}} dz = p Q\left[\sqrt{\frac{2E_b}{N_0}}\right] + (1-p) Q\left[\sqrt{\frac{2E_b}{N_0}}\right] = Q\left[\sqrt{\frac{2E_b}{N_0}}\right]$$

5. a. Select s_1 & normalize

$$\mathcal{E}_1 = \frac{A^2 T}{3} + \frac{2A^2 T}{3} = A^2 T$$

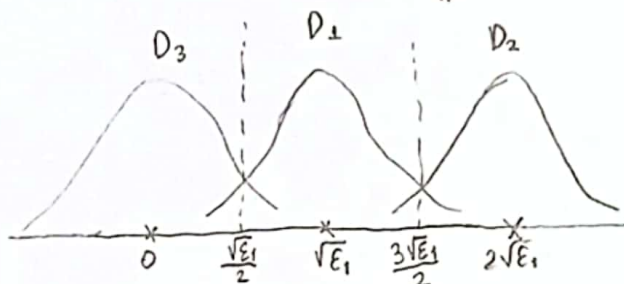
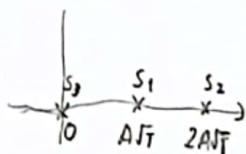
$$\phi_1(t) = \frac{s_1(t)}{\sqrt{\mathcal{E}_1}} = \begin{cases} \frac{1}{\sqrt{T}} & 0 \leq t \leq \frac{T}{3} \\ -\frac{1}{\sqrt{T}} & \frac{T}{3} \leq t \leq T \end{cases}$$



$$s_1 = \sqrt{\mathcal{E}_1} = A\sqrt{T}$$

$$s_2 = 2\sqrt{\mathcal{E}_1} = 2A\sqrt{T}$$

$$s_3 = 0$$



b) Since equiprobable

$$D_3 = \{r \in R : f(r|s_3) > f(r|s_1), f(r|s_3) > f(r|s_2)\} = \left\{r \in R : r < \frac{\sqrt{\mathcal{E}_1}}{2}\right\}$$

$$D_1 = \{r \in R : f(r|s_1) > f(r|s_2), f(r|s_1) > f(r|s_3)\} = \left\{r \in R : \frac{\sqrt{\mathcal{E}_1}}{2} < r < \frac{3\sqrt{\mathcal{E}_1}}{2}\right\}$$

$$D_2 = \{r \in R : f(r|s_2) > f(r|s_1), f(r|s_2) > f(r|s_3)\} = \left\{r \in R : r > \frac{3\sqrt{\mathcal{E}_1}}{2}\right\}$$

$$c) P_e = \sum_{m=1}^M p_m \left(\sum_{\substack{1 \leq m' \leq M \\ m' \neq m}} \int_{D_{m'}} (p(r|s_m) dr) \right)$$

$$P_e = \frac{1}{3} \int_{D_3} p(r|s_1) dr + \frac{1}{3} \int_{D_2} p(r|s_1) dr + \frac{1}{3} \left(\int_{D_1} p(r|s_3) dr + \int_{D_2} p(r|s_3) dr \right) + \frac{1}{3} \left(\int_{D_3} p(r|s_2) dr + \int_{D_1} p(r|s_2) dr \right)$$

$$\frac{1}{3} Q \left[\sqrt{\frac{\mathcal{E}_1}{2N_0}} \right] + \frac{1}{3} Q \left[\sqrt{\frac{\mathcal{E}_1}{2N_0}} \right] + \frac{1}{3} Q \left[\sqrt{\frac{\mathcal{E}_1}{2N_0}} \right] + \frac{1}{3} Q \left[\sqrt{\frac{\mathcal{E}_1}{2N_0}} \right]$$

$$P_e = \frac{4}{3} Q \left[\sqrt{\frac{\mathcal{E}_1}{2N_0}} \right] \rightarrow \text{we need it in terms of SNR:}$$

$$\mathcal{E}_{avg} = \frac{1}{3} (0 + A^2 T + 4A^2 T)$$

$$\mathcal{E}_{avg} = \frac{1}{3} (0 + \mathcal{E}_1 + 4\mathcal{E}_1) = \frac{5\mathcal{E}_1}{3}$$

$$\mathcal{E}_{b,avg} = \frac{\mathcal{E}_{avg}}{\log_2 M} = \frac{5\mathcal{E}_1}{3 \log_2 3} \Rightarrow \mathcal{E}_1 = 0.95 \mathcal{E}_{b,avg}$$

$$P_e = \frac{4}{3} Q \left[\frac{0.95 \mathcal{E}_{b,avg}}{2N_0} \right]$$

$$d) R = R, \log_2 M = 3000 \log_2 3$$

Q6 a) Structure of optimal receiver
 optimum receiver metric to maximize

$$f(\vec{r}|\vec{s}_m) = \prod_{k=1}^N f(r_k|s_{mk})$$

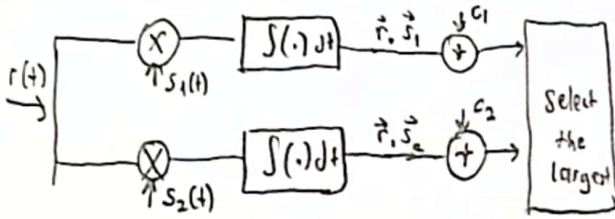
Do not write

Vector form:

$$C(\vec{r}, \vec{s}_m) = \vec{r} \cdot \vec{s}_m - \frac{1}{2} |\vec{s}_m|^2 + \frac{N_0}{2} \ln P(s_m)$$

$$C(r, s_m) = \int_{-\infty}^{\infty} r(t) s_m(t) dt - \frac{1}{2} \int_{-\infty}^{\infty} |s_m(t)|^2 dt + \frac{N_0}{2} \ln P(s_m)$$

correlator bias term



Since $s_1(t) = -s_2(t)$ energy of the signals are same

$$\int_{-\infty}^{\infty} r(t) s_1(t) dt + \frac{N_0}{2} \ln P(s_1) \underset{s_2}{\overset{s_1}{>}} \int_{-\infty}^{\infty} r(t) s_2(t) dt + \frac{N_0}{2} \ln P(s_2)$$

$$\int_{-\infty}^{\infty} r(t) s_1(t) dt \underset{s_2}{\overset{s_1}{>}} \frac{N_0}{4} \ln \frac{P_2}{P_1}$$

b) $P_e = ?$

first $P(e|s_1)$

$$\int_{-\infty}^{\infty} (s_1(t) + n(t)) (s_2(t)) dt = \int_0^T (s_1(t))^2 dt + \int_0^T n(t) s_1(t) dt$$

$\epsilon_s + n \rightarrow$ zero mean gaussian with variance

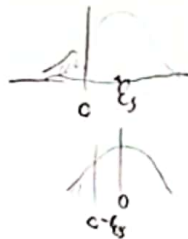
$$\sigma_n^2 = E \left[\int_0^T n(z) s_1(z) dz \int_0^T n(v) s_1(v) dv \right]$$

$$= \int_0^T \int_0^T s_1(z) s_1(v) E[n(z) n(v)] dz dv$$

$\rightarrow \frac{N_0}{2} \delta(z-v)$

$$= \frac{N_0}{2} \int_0^T (s_1(z))^2 dz$$

$$\sigma_n^2 = \frac{N_0}{2} \epsilon_s$$

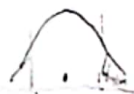


$$P(e|s_1) = \int_{-\infty}^{\infty} \frac{1}{\sqrt{\pi N_0 \epsilon_s}} e^{-\frac{x^2}{N_0 \epsilon_s}} dx$$

$$r = \sqrt{\frac{N_0 \epsilon_s}{2}} z$$

$$z = \left(\frac{N_0}{4} \ln \frac{P_2}{P_1} - \epsilon_s \right) / \sqrt{\frac{N_0 \epsilon_s}{2}}$$

$$Q \left(\frac{\epsilon_s \sqrt{2}}{\sqrt{N_0 \epsilon_s}} - \frac{\sqrt{N_0}}{2\sqrt{2}} \ln \frac{P_2}{P_1} \right)$$



$$P(e|s_1) = Q \left(\sqrt{\frac{2\epsilon_s}{N_0}} - \frac{\sqrt{N_0}}{2\sqrt{2}} \ln \frac{P_2}{P_1} \right)$$

$$P(e) = P_1 Q \left(\sqrt{\frac{2\epsilon_s}{N_0}} - \frac{\sqrt{N_0}}{2\sqrt{2}} \ln \frac{P_2}{P_1} \right) + (1-P_1) Q \left(\sqrt{\frac{2\epsilon_s}{N_0}} + \frac{\sqrt{N_0}}{2\sqrt{2}} \ln \frac{P_2}{P_1} \right)$$

Q.7) Equiprobable binary antipodal (as we did in Q4)

a)



$$r = n : \sqrt{E_b}$$

$$P(e) = P(e|s_0)P(s_0) + P(e|s_1)P(s_1)$$

$$P_e = \frac{1}{2} \int_0^{\infty} f(r|s_0) dr + \frac{1}{2} \int_{-\infty}^0 f(r|s_1) dr$$

$$P_e = \int_0^{\infty} e^{-\frac{r^2}{2}} dr$$

this part is solved in Q4.b

$$\sqrt{\frac{N_0}{2}} z - \sqrt{E_b} = 0$$

$$P_e = Q\left[\sqrt{\frac{2E_b}{N_0}}\right]$$

$$z = \frac{x - \mu}{\sigma}$$

$$x = \sigma z + \mu$$

$$r = \sqrt{\frac{N_0}{2}} z - \sqrt{E_b}$$

b) Optimal Detector does not change for N_0 or N_1 decision region is at 0 regardless

$$P_1 = Q\left[\sqrt{\frac{2E_b}{N_1}}\right] \quad P_1 > P_e$$

c) same, does not effect the detector so $P_{e1} = Q\left[\sqrt{\frac{2E_b}{N_1}}\right] = P_1$

d)

Decision region

$$\frac{p_0}{\sqrt{\pi N_0}} e^{-\frac{(r - \sqrt{E_b})^2}{N_0}} = \frac{p_1}{\sqrt{\pi N_0}} e^{-\frac{(r + \sqrt{E_b})^2}{N_0}}$$

$$P(s_0) p(r|s_0) = P(s_1) p(r|s_1)$$

$$e^{\frac{(r - \sqrt{E_b})^2}{N_0} - \frac{(r + \sqrt{E_b})^2}{N_0}}$$

$$e^{\frac{4r\sqrt{E_b}}{N_0}} = \frac{p_1}{p_0}$$

$$r_{th} = \frac{N_0}{4\sqrt{E_b}} \ln \frac{(1-p)}{p} \rightarrow \text{threshold depends on } N_0$$

$$P_1 > P_{e1} > P_e$$

↓ due to sub optimal solution
↓ due to higher noise level

Q.8) a) Assume we use matched filter

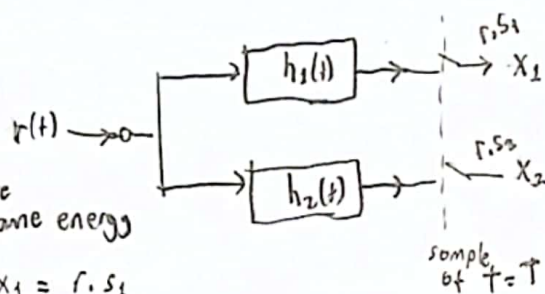
$$h_1(t) = s_1(T-t)$$

$$h_2(t) = s_2(T-t)$$

since signals are equiprobable & have same energy

$$x_1 \geq x_2$$

where $x_1 = r \cdot s_1$
 $x_2 = r \cdot s_2$



$$\hat{m} = \arg \max [\eta_m + r \cdot s_m]$$

where $\eta_m = \frac{N_0}{2} \ln P_m - \frac{1}{2} E_m$

you can discard these terms if equiprobable energy

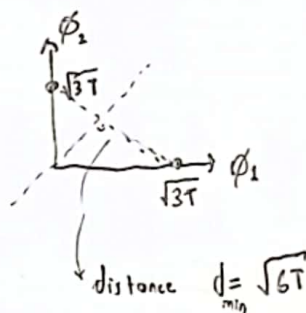
b) $P(e) = ?$ if $\frac{E}{N_0} = 4$

$$E_2 = E_1 = \int_0^T (1)^2 dt + \int_T^{2T} (-1)^2 dt = 3T = E \quad \text{given as signal energy}$$

$$\phi_1(t) = \frac{s_1(t)}{\sqrt{3T}}$$

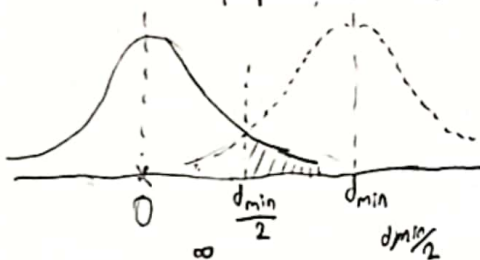
$$\phi_2(t) = \frac{s_2(t)}{\sqrt{3T}}$$

since orthogonal



binary &

Since equiprobable: we can regard this as:



$$P_e = \frac{1}{2} \int_{\frac{d_{\min}}{2}}^{\infty} f(r|s_1) dr + \frac{1}{2} \int_{-\infty}^{\frac{d_{\min}}{2}} f(r|s_2) dr$$

$$P_e = \int_{\frac{d_{\min}}{2}}^{\infty} f(r|s_1) dr =$$

$$P_e = \int_{\frac{d_{\min}}{2} / \sqrt{\frac{N_0}{2}}}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}} dz = Q \left[\frac{d_{\min}/2}{\sqrt{N_0/2}} \right] = Q \left[\sqrt{\frac{3T}{N_0}} \right] = Q \left[\sqrt{\frac{E}{N_0}} \right]$$

Note: $Q(x)$ can be denoted as $\frac{1}{2} \operatorname{erfc} \left(\frac{x}{\sqrt{2}} \right)$ in some solutions.

Q.9 | Optimum detection of sinusoidal signal $s(t) = \sin\left(\frac{8\pi t}{T}\right)$ $0 \leq t \leq T$

a) Determine correlator output of noiseless input

$$y(T) = \int_0^T r(z) s(z) dz$$

$$y(T) = \int_0^T s^2(z) dz = \int_0^T \sin^2\left(\frac{8\pi z}{T}\right) dz = \int_0^T \frac{1}{2} \left[1 - \cos\left(\frac{16\pi z}{T}\right) \right] dz$$

$$\boxed{\frac{T}{2}}$$

b) Determine the corresponding matched filter

$$h(t) = s(T-t)$$

$$y(t) = \int_{-\infty}^{\infty} r(\lambda) h(t-\lambda) d\lambda$$

$$\sin x \cdot \sin y = -\frac{1}{2} [\cos(x+y) - \cos(x-y)]$$

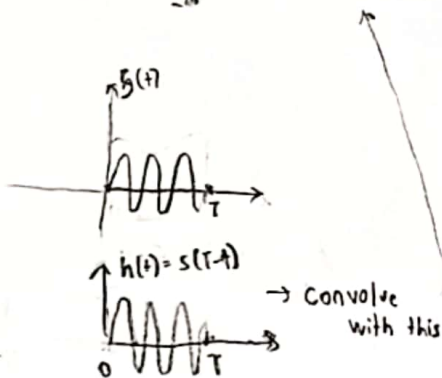
$$y(t) = \int_{-\infty}^{\infty} s(\lambda) s(T-t+\lambda) d\lambda = \int_{-\infty}^{\infty} \sin\left(\frac{8\pi \lambda}{T}\right) \sin\left(\frac{8\pi (T-t+\lambda)}{T}\right) d\lambda$$

$$= \frac{1}{2} \int_{-\infty}^{\infty} \cos\left[\frac{8\pi (T-t)}{T}\right] d\lambda - \frac{1}{2} \int_{-\infty}^{\infty} \cos\left[\frac{8\pi (T-t+2\lambda)}{T}\right] d\lambda$$

$$= \frac{1}{2} \left(\cos\left[\frac{8\pi (T-t)}{T}\right] \lambda \right) \Big|_0^T - \frac{T}{32\pi} \sin\left[\frac{8\pi (T-t+2\lambda)}{T}\right] \Big|_{\lambda=0}^{\lambda=T}$$

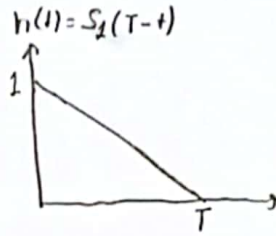
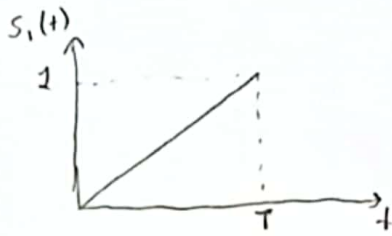
$$= \frac{T}{2} \cos\left[\frac{8\pi (T-t)}{T}\right] - \frac{T}{32\pi} \sin\left[\frac{8\pi (3T-t)}{T}\right] + \frac{T}{32\pi} \sin\left[\frac{8\pi (T-t)}{T}\right]$$

at $t=T$ $\boxed{y(T) = \frac{T}{2}}$



→ Results are the same

Q 10.

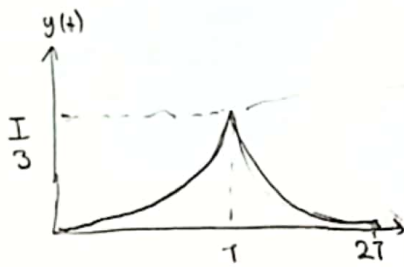


$$s_1(z) = \begin{cases} z/T & 0 < z < T \\ 0 & \text{o.w.} \end{cases}$$

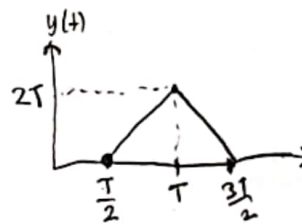
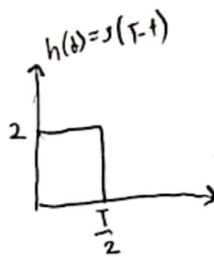
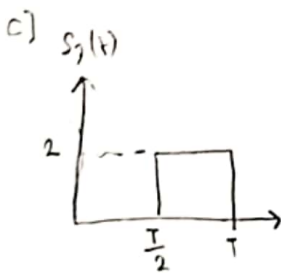
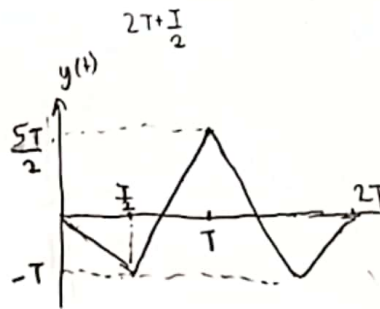
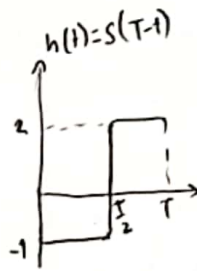
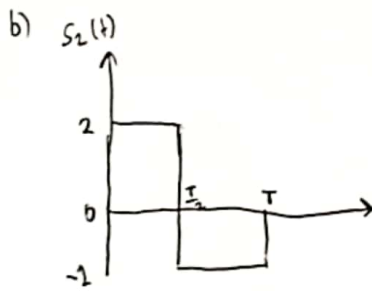
$$h(z) = \begin{cases} 1 - \frac{z}{T} & 0 < z < T \\ 0 & \text{o.w.} \end{cases}$$

$$y_1(t) = \int_{-\infty}^{\infty} s_1(z) h(t-z) dz$$

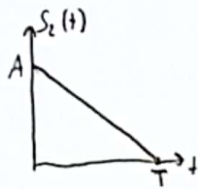
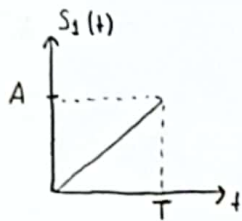
$$y_1(t) = \int_0^T \frac{z}{T} \left(1 - \frac{(t-z)}{T} \right) dz$$



$$\int_0^T \frac{t^2}{T^2} dt = \frac{T}{3}$$



11.



→ Two signals have the same energy

- a) Structure of the optimal receiver → equiprobable
→ same energy so no bias term

$$\int_{-\infty}^{\infty} r(t) s_1(t) dt \gtrless_{s_2} \int_{-\infty}^{\infty} r(t) s_2(t) dt$$

- b) $P(e) = ?$ first try to find $P(e|s_1)$

$$\int_{-\infty}^{\infty} s_1(t) (s_1(t) - s_2(t)) dt + \int_{-\infty}^{\infty} n(t) (s_1(t) - s_2(t)) dt \gtrless_{s_2} 0$$

$$w = \frac{A^2 T}{6}$$

+ n

$$\begin{aligned} \sigma_n^2 &= E \left[\int_{-\infty}^{\infty} n(t) (s_1(t) - s_2(t)) dt \int_{-\infty}^{\infty} n(v) (s_1(v) - s_2(v)) dv \right] \\ &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} (s_1(t) - s_2(t)) (s_1(v) - s_2(v)) \frac{N_0}{2} \delta(v-t) dv dt \\ &= \frac{N_0}{2} \int_{-\infty}^{\infty} (s_1(t) - s_2(t))^2 dt \\ &= \int_0^T \left(\frac{2A\tau}{T} - A \right)^2 d\tau \\ &= \frac{N_0 A^2 T}{6} \end{aligned}$$

$$P(e|s_1) = \int_{-\infty}^0 f(w) dw, \quad w \text{ has } \frac{A^2 T}{6} \text{ mean and } \sqrt{\frac{N_0 A^2 T}{6}} \text{ standard deviation}$$

$$r = \sqrt{\frac{N_0 A^2 T}{6}} z + \frac{A^2 T}{6}$$

$$r = \sigma z + \mu$$

$$P(e|s_1) = \int_{-\infty}^{-\frac{A^2 T}{6} / \sqrt{\frac{N_0 A^2 T}{6}}} \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}} dz = Q \left[\sqrt{\frac{A^2 T}{6 N_0}} \right] = P(e|s_2)$$

$$P_e = \frac{1}{2} Q \left[\sqrt{\frac{A^2 T}{6 N_0}} \right] + \frac{1}{2} Q \left[\sqrt{\frac{A^2 T}{6 N_0}} \right] = Q \left[\sqrt{\frac{A^2 T}{6 N_0}} \right]$$

$$12. \quad r = \underset{\substack{\downarrow \\ \text{equally} \\ \text{probable}}}{\pm A} + \underset{\substack{\downarrow \\ \text{Laplacian}}}{n}$$

$$p(n) = \frac{1}{\sqrt{2\sigma}} e^{-\frac{|n|\sqrt{2}}{\sigma}} \quad \lambda = \frac{\sqrt{2}}{\sigma}$$

$$f(n) = \frac{\lambda}{2} e^{-\lambda|n|}$$

Optimal Receiver

$$\frac{f(r|A)}{f(r|-A)} = e^{-\lambda[|r-A|-|r+A|]} \underset{-A}{\overset{A}{\gtrless}} 1$$

$$-\lambda[|r-A|-|r+A|] \underset{-A}{\overset{A}{\gtrless}} 0 \Rightarrow r \underset{-A}{\overset{A}{\gtrless}} 0$$



$$P(e) = \frac{1}{2} P(e|A) + \frac{1}{2} P(e|-A)$$

$$\frac{1}{2} \int_{-\infty}^0 f(r|A) dr + \frac{1}{2} \int_0^{\infty} f(r|-A) dr$$

$$= \frac{1}{2} \int_{-\infty}^0 \frac{\lambda}{2} e^{-\lambda|r-A|} dr + \frac{1}{2} \int_0^{\infty} \frac{\lambda}{2} e^{-\lambda|r+A|} dr$$

$$= \frac{\lambda}{2} \int_A^{\infty} e^{-\lambda|x|} dx \quad \swarrow \text{symmetric}$$

$$= -\frac{1}{2} e^{-\lambda x} \Big|_A^{\infty} = \frac{1}{2} e^{-\lambda A}$$