

otherwise it decides in favor of s_1 . The decision rule may be expressed as:

$$\frac{\text{PM}(\mathbf{r}, \mathbf{s}_0)}{\text{PM}(\mathbf{r}, \mathbf{s}_1)} = e^{\frac{(r-A\sqrt{T})^2 - r^2}{N_0}} = e^{-\frac{(2r-A\sqrt{T})A\sqrt{T}}{N_0}} \begin{array}{l} s_0 \\ \geq \\ < \\ s_1 \end{array} \quad 1$$

or equivalently :

$$r \begin{array}{l} s_1 \\ \geq \\ < \\ s_0 \end{array} \frac{1}{2}A\sqrt{T}$$

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

(b) The average probability of error is:

$$\begin{aligned} P(e) &= \frac{1}{2}P(e|s_0) + \frac{1}{2}P(e|s_1) \\ &= \frac{1}{2} \int_{\frac{1}{2}A\sqrt{T}}^{\infty} p(r|s_0)dr + \frac{1}{2} \int_{-\infty}^{\frac{1}{2}A\sqrt{T}} p(r|s_1)dr \\ &= \frac{1}{2} \int_{\frac{1}{2}A\sqrt{T}}^{\infty} \frac{1}{\sqrt{\pi N_0}} e^{-\frac{r^2}{N_0}} dr + \frac{1}{2} \int_{-\infty}^{\frac{1}{2}A\sqrt{T}} \frac{1}{\sqrt{\pi N_0}} e^{-\frac{(r-A\sqrt{T})^2}{N_0}} dr \\ &= \frac{1}{2} \int_{\frac{1}{2}\sqrt{\frac{2}{N_0}}A\sqrt{T}}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx + \frac{1}{2} \int_{-\infty}^{-\frac{1}{2}\sqrt{\frac{2}{N_0}}A\sqrt{T}} \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx \\ &= Q \left[\frac{1}{2} \sqrt{\frac{2}{N_0}} A\sqrt{T} \right] = Q \left[\sqrt{\text{SNR}} \right] \end{aligned}$$

where

$$\text{SNR} = \frac{\frac{1}{2}A^2T}{N_0}$$

Thus, the on-off signaling requires a factor of two more energy to achieve the same probability of error as the antipodal signaling.

Problem 5.5 :

Since $\{f_n(t)\}$ constitute an orthonormal basis for the signal space : $r(t) = \sum_{n=1}^N r_n f_n(t)$, $s_m(t) =$

$\sum_{n=1}^N s_{mn} f_n(t)$. Hence, for any m :

$$\begin{aligned}
C(\mathbf{r}, \mathbf{s}_m) &= 2 \int_0^T r(t) s_m(t) dt - \int_0^T s_m^2(t) dt \\
&= 2 \int_0^T \sum_{n=1}^N r_n f_n(t) \sum_{l=1}^N s_{ml} f_l(t) dt - \int_0^T \sum_{n=1}^N s_{mn} f_n(t) \sum_{l=1}^N s_{ml} f_l(t) dt \\
&= 2 \sum_{n=1}^N r_n \sum_{l=1}^N s_{ml} \int_0^T f_n(t) f_l(t) dt - \sum_{n=1}^N s_{mn} \sum_{l=1}^N s_{ml} \int_0^T f_n(t) f_l(t) dt \\
&= 2 \sum_{n=1}^N r_n s_{mn} - \sum_{n=1}^N s_{mn}^2
\end{aligned}$$

where we have exploited the orthonormality of $\{f_n(t)\}$: $\int_0^T f_n(t) f_l(t) dt = \delta_{nl}$. The last form is indeed the original form of the correlation metrics $C(\mathbf{r}, \mathbf{s}_m)$.

Problem 5.6 :

The SNR at the filter output will be :

$$SNR = \frac{|y(T)|^2}{E[|n(T)|^2]}$$

where $y(t)$ is the part of the filter output that is due to the signal $s_l(t)$, and $n(t)$ is the part due to the noise $z(t)$. The denominator is :

$$\begin{aligned}
E[|n(T)|^2] &= \int_0^T \int_0^T E[z(a)z^*(b)] h_l(T-a)h_l^*(T-b) da db \\
&= 2N_0 \int_0^T |h_l(T-t)|^2 dt
\end{aligned}$$

so we want to maximize :

$$SNR = \frac{\left| \int_0^T s_l(t) h_l(T-t) dt \right|^2}{2N_0 \int_0^T |h_l(T-t)|^2 dt}$$

From Schwartz inequality :

$$\left| \int_0^T s_l(t) h_l(T-t) dt \right|^2 \leq \int_0^T |h_l(T-t)|^2 dt \int_0^T |s_l(t)|^2 dt$$

Hence :

$$SNR \leq \frac{1}{2N_0} \int_0^T |s_l(t)|^2 dt = \frac{\mathcal{E}}{N_0} = SNR_{\max}$$

and the maximum occurs when :

$$s_l(t) = h_l^*(T-t) \Leftrightarrow h_l(t) = s_l^*(T-t)$$

Problem 5.7 :

$$N_{mr} = \text{Re} \left[\int_0^T z(t) f_m^*(t) dt \right]$$

(a) Define $a_m = \int_0^T z(t) f_m^*(t) dt$. Then, $N_{mr} = \text{Re}(a_m) = \frac{1}{2} [a_m + a_m^*]$.

$$E(N_{mr}) = \text{Re} \left[\int_0^T E(z(t)) f_m^*(t) dt \right] = 0$$

since, $E[z(t)] = 0$. Also :

$$E(N_{mr}^2) = E \left[\frac{a_m^2 + (a_m^*)^2 + 2a_m a_m^*}{4} \right]$$

But $E(a_m^2) = E \left[\int_0^T \int_0^T z(a) z(b) f_m^*(a) f_m^*(b) da db \right] = 0$, since $E[z(a)z(b)] = 0$ (Problem 4.3), and the same is true for $E[(a_m^*)^2] = 0$, since $E[z^*(a)z^*(b)] = 0$ Hence :

$$\begin{aligned} E(N_{mr}^2) &= E \left[\frac{a_m a_m^*}{2} \right] = \frac{1}{2} \int_0^T \int_0^T E[z(a)z^*(b)] f_m^*(a) f_m(b) da db \\ &= N_0 \int_0^T |f_m(a)|^2 da = 2\mathcal{E}N_0 \end{aligned}$$

(b) For $m \neq k$:

$$\begin{aligned} E[N_{mr}N_{kr}] &= E \left[\frac{a_m + a_m^*}{2} \frac{a_k + a_k^*}{2} \right] \\ &= E \left[\frac{a_m a_k + a_m^* a_k^* + a_m a_k^* + a_m^* a_k}{4} \right] \end{aligned}$$

But, similarly to part (a), $E[a_m a_k] = E[a_m^* a_k^*] = 0$, hence, $E[N_{mr}N_{kr}] = E \left[\frac{a_m^* a_k + a_m a_k^*}{4} \right]$. Now :

$$\begin{aligned} E[a_m a_k^*] &= \int_0^T \int_0^T E[z(a)z^*(b)] f_m^*(a) f_k(b) da db \\ &= 2N_0 \int_0^T f_m^*(a) f_k(a) da = 0 \end{aligned}$$

since, for $m \neq k$, the waveforms are orthogonal.

Similarly : $E[a_m^* a_k] = 0$, hence : $E[N_{mr}N_{kr}] = 0$.

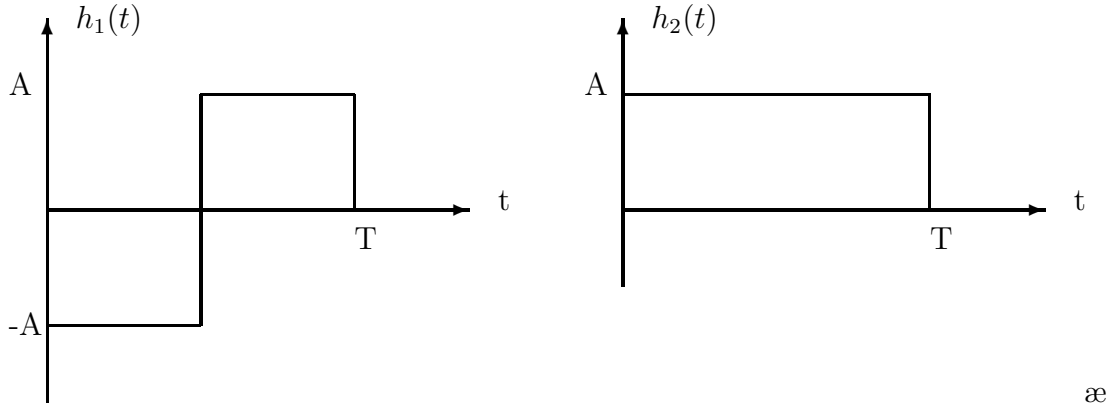
Problem 5.8 :

(a) Since the given waveforms are the equivalent lowpass signals :

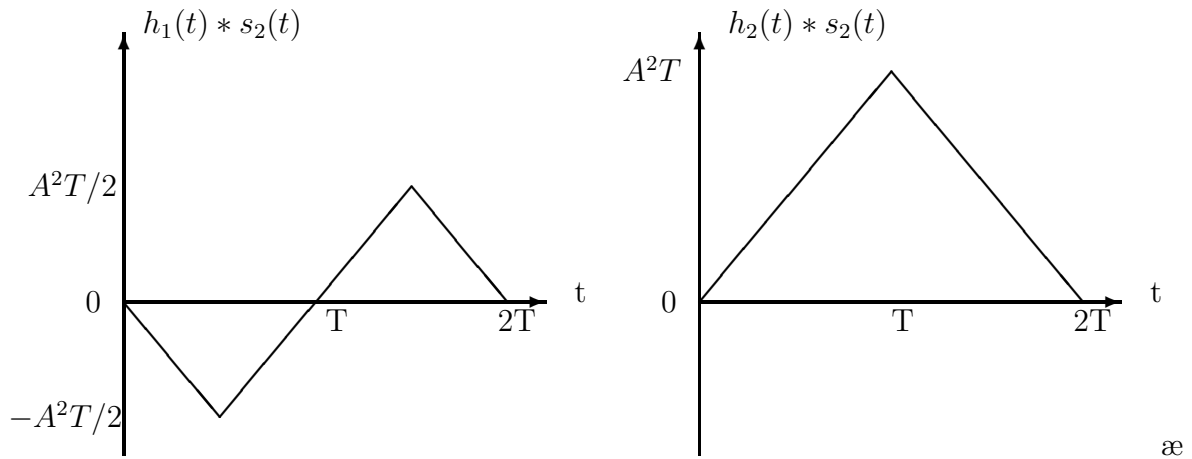
$$\begin{aligned} \mathcal{E}_1 &= \frac{1}{2} \int_0^T |s_1(t)|^2 dt = \frac{1}{2} A^2 \int_0^T dt = A^2 T / 2 \\ \mathcal{E}_2 &= \frac{1}{2} \int_0^T |s_2(t)|^2 dt = \frac{1}{2} A^2 \int_0^T dt = A^2 T / 2 \end{aligned}$$

Hence $\mathcal{E}_1 = \mathcal{E}_2 = \mathcal{E}$. Also $\rho_{12} = \frac{1}{2\mathcal{E}} \int_0^T s_1(t)s_2^*(t)dt = 0$.

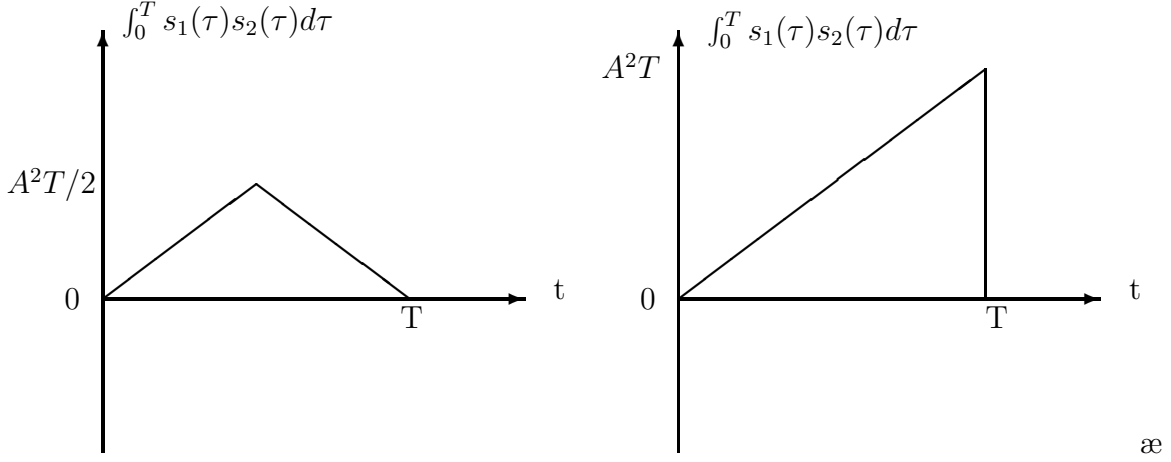
(b) Each matched filter has an equivalent lowpass impulse response : $h_i(t) = s_i(T - t)$. The following figure shows $h_i(t)$:



(c)



(d)



(e) The outputs of the matched filters are different from the outputs of the correlators. The two sets of outputs agree at the sampling time $t = T$.

(f) Since the signals are orthogonal ($\rho_{12} = 0$) the error probability for AWGN is $P_2 = Q\left(\sqrt{\frac{\mathcal{E}}{N_0}}\right)$, where $\mathcal{E} = A^2T/2$.

Problem 5.9 :

(a) The joint pdf of a, b is :

$$p_{ab}(a, b) = p_{xy}(a - m_r, b - m_i) = p_x(a - m_r)p_y(b - m_i) = \frac{1}{2\pi\sigma^2}e^{-\frac{1}{2\sigma^2}[(a-m_r)^2+(b-m_i)^2]}$$

(b) $u = \sqrt{a^2 + b^2}$, $\phi = \tan^{-1}b/a \Rightarrow a = u \cos \phi$, $b = u \sin \phi$ The Jacobian of the transformation is : $J(a, b) = \begin{vmatrix} \partial a/\partial u & \partial a/\partial \phi \\ \partial b/\partial u & \partial b/\partial \phi \end{vmatrix} = u$, hence :

$$\begin{aligned} p_{u\phi}(u, \phi) &= \frac{u}{2\pi\sigma^2}e^{-\frac{1}{2\sigma^2}[(u \cos \phi - m_r)^2 + (u \sin \phi - m_i)^2]} \\ &= \frac{u}{2\pi\sigma^2}e^{-\frac{1}{2\sigma^2}[u^2 + M^2 - 2uM \cos(\phi - \theta)]} \end{aligned}$$

where we have used the transformation :

$$\left\{ \begin{array}{l} M = \sqrt{m_r^2 + m_i^2} \\ \theta = \tan^{-1}m_i/m_r \end{array} \right\} \Rightarrow \left\{ \begin{array}{l} m_r = M \cos \theta \\ m_i = M \sin \theta \end{array} \right\}$$