

ÉCOLE POLYTECHNIQUE FÉDÉRALE DE LAUSANNE

School of Computer and Communication Sciences

Handout 22

Principles of Digital Communications

Solutions to Problem Set 9

May 5, 2015

SOLUTION 1.

(a)

$$\begin{aligned} R_\xi(\tau) &= \int_{-\infty}^{\infty} \xi(t+\tau)\xi^*(t) dt = \langle \xi(t+\tau), \xi(t) \rangle \\ &\stackrel{(1)}{\leq} \|\xi(t+\tau)\| \cdot \|\xi(t)\| = \|\xi\| \cdot \|\xi\| = \|\xi\|^2 \stackrel{(2)}{=} R_\xi(0), \end{aligned}$$

where (1) follows from the Cauchy-Schwarz inequality and (2) from the fact that $R_\xi(0) = \int_{-\infty}^{\infty} \xi(t)\xi^*(t) dt = \|\xi\|^2$.

(b)

$$\begin{aligned} R_\xi(-\tau) &= \int_{-\infty}^{\infty} \xi(t-\tau)\xi^*(t) dt \\ &= \left(\int_{-\infty}^{\infty} \xi(t)\xi^*(t-\tau) dt \right)^* \\ &\stackrel{t \rightarrow t+\tau}{=} R_\xi^*(\tau). \end{aligned}$$

(c)

$$\begin{aligned} R_\xi(\tau) &= \int_{-\infty}^{\infty} \xi(t+\tau)\xi^*(t) dt \\ &\stackrel{t \rightarrow t-\tau}{=} \int_{-\infty}^{\infty} \xi(t)\xi^*(t-\tau) dt \\ &= \xi(\tau) \star \xi^*(-\tau). \end{aligned}$$

(d) By Parseval's identity, we have

$$\begin{aligned} R_\xi(\tau) &= \langle \xi(t+\tau), \xi(t) \rangle \\ &= \langle \xi_{\mathcal{F}}(f)e^{j2\pi f\tau}, \xi_{\mathcal{F}}(f) \rangle \\ &= \int_{-\infty}^{\infty} \xi_{\mathcal{F}}(f)\xi_{\mathcal{F}}^*(f)e^{j2\pi f\tau} df \\ &= \int_{-\infty}^{\infty} |\xi_{\mathcal{F}}(f)|^2 e^{j2\pi f\tau} df, \end{aligned}$$

which is the inverse Fourier transform of $|\xi_{\mathcal{F}}(f)|^2$.

SOLUTION 2.

(a) We have

$$y(t) = \int_{-\infty}^{\infty} w(\tau)\psi(\tau - t)d\tau.$$

The samples of this waveform at multiples of T are

$$\begin{aligned} y(mT) &= \int_{-\infty}^{\infty} w(\tau)\psi(\tau - mT)d\tau \\ &= \int_{-\infty}^{\infty} \left[\sum_{k=1}^K d_k \psi(\tau - kT) \right] \psi(\tau - mT)d\tau \\ &= \sum_{k=1}^K d_k \int_{-\infty}^{\infty} \psi(\tau - kT)\psi(\tau - mT)d\tau \\ &= \sum_{k=1}^K d_k \mathbb{1}\{k = m\} \\ &= d_m. \end{aligned}$$

(b) Let $\tilde{w}(t)$ be the channel output. Then, $\tilde{y}(t)$ is $\tilde{w}(t)$ filtered by $\psi(-t)$. We have

$$\tilde{w}(t) = w(t) + \rho w(t - T)$$

and

$$\tilde{y}(t) = \int_{-\infty}^{\infty} \tilde{w}(\tau)\psi(\tau - t)d\tau.$$

The samples of this waveform at multiples of T are

$$\begin{aligned} \tilde{y}(mT) &= \int_{-\infty}^{\infty} \tilde{w}(\tau)\psi(\tau - mT)d\tau \\ &= \int_{-\infty}^{\infty} [w(\tau) + \rho w(\tau - T)]\psi(\tau - mT)d\tau \\ &= \int_{-\infty}^{\infty} \left[\sum_{k=1}^K d_k \psi(\tau - kT) \right] \psi(\tau - mT)d\tau + \\ &\quad \rho \int_{-\infty}^{\infty} \left[\sum_{k=1}^K d_k \psi(\tau - T - kT) \right] \psi(\tau - mT)d\tau \\ &= \sum_{k=1}^K d_k \mathbb{1}\{k = m\} + \rho \sum_{k=1}^K d_k \mathbb{1}\{k = m - 1\} \\ &= d_m + \rho d_{m-1}. \end{aligned}$$

(c) From the symmetry of the problem, we have

$$P_e = P_e(1) = P_e(-1).$$

$$\begin{aligned} P_e(1) &= \Pr \left\{ \hat{D}_k = -1 | D_k = 1, D_{k-1} = -1 \right\} \Pr \{ D_{k-1} = -1 \} + \\ &\quad \Pr \left\{ \hat{D}_k = -1 | D_k = 1, D_{k-1} = 1 \right\} \Pr \{ D_{k-1} = 1 \} \\ &= \frac{1}{2} (\Pr \{ Y_k < 0 | D_k = 1, D_{k-1} = -1 \} + \Pr \{ Y_k < 0 | D_k = 1, D_{k-1} = 1 \}) \\ &= \frac{1}{2} (\Pr \{ 1 - \alpha + Z_k < 0 \} + \Pr \{ 1 + \alpha + Z_k < 0 \}) \\ &= \frac{1}{2} (\Pr \{ Z_k < -1 + \alpha \} + \Pr \{ Z_k < -1 - \alpha \}) \\ &= \frac{1}{2} \left[Q \left(\frac{1 - \alpha}{\sigma} \right) + Q \left(\frac{1 + \alpha}{\sigma} \right) \right]. \end{aligned}$$

SOLUTION 3.

(a) We can easily see that

$$\mathbb{E} [X_i | X_{i-1}] = \frac{1}{2} X_{i-1} + \frac{1}{2} (-X_{i-1}) = 0.$$

Consequently (using the law of total expectation)

$$\mathbb{E} [X_i] = \mathbb{E} [\mathbb{E} [X_i | X_{i-1}]] = 0.$$

Therefore,

$$K_X[k] = \mathbb{E} [(X_i - \mathbb{E} [X_i])(X_{i-k} - \mathbb{E} [X_{i-k}])^*] = \mathbb{E} [X_i X_{i-k}^*]$$

Moreover, using the fact that $X_i = X_{i-1} \times (-1)^{D_i}$ repeatedly, we can write

$$X_i = X_{i-k} \times \prod_{j=i-k+1}^i (-1)^{D_j}$$

Thus,

$$\begin{aligned} K_X[k] &= \mathbb{E} [X_i X_{i-k}^*] \\ &= \mathbb{E} \left[X_{i-k} \prod_{j=i-k+1}^i (-1)^{D_j} X_{i-k}^* \right] \\ &\stackrel{(*)}{=} \mathbb{E} [X_{i-k} X_{i-k}^*] \prod_{j=i-k+1}^i \mathbb{E} [(-1)^{D_j}] \\ &= \mathcal{E} \prod_{j=i-k+1}^i \mathbb{E} [(-1)^{D_j}] \\ &= \begin{cases} \mathcal{E} & \text{if } k = 0, \\ 0 & \text{otherwise.} \end{cases} \end{aligned}$$

where (\star) follows since the data D_i are independent. We have also used the fact that $\mathbb{E} [(-1)^{D_i}] = 0$.

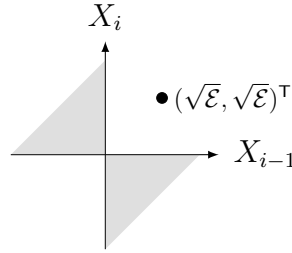
- (b) By sampling the signal at the output of the matched filter, $Y(t)$, at multiples of T , we obtain

$$Y(iT) = X_i + Z_i,$$

where Z_i is normally distributed with zero mean and variance $N_0/2$. By looking at the definition of X_i , we see that it is equal to X_{i-1} if $D_i = 0$ and equal to $-X_{i-1}$ if $D_i = 1$. Therefore a simple decoder estimates that $\hat{D}_i = 0$ if Y_i and Y_{i-1} have the same sign, and $\hat{D}_i = 1$ otherwise. This is equivalent to

$$Y_i Y_{i-1} \underset{\hat{D}_i=1}{\overset{\hat{D}_i=0}{\geq}} 0.$$

- (c) We first compute the error probability when $D_i = 0$. If $X_{i-1} = \sqrt{\mathcal{E}}$, then $X_i = \sqrt{\mathcal{E}}$. When we decode, we will make an error if the signal $(Y_{i-1}, Y_i)^\top$ is in the second or fourth quadrants (shaded regions in the following figure).



Due to the symmetry of the problem, the probability for this to happen is two times the probability for $(Y_{i-1}, Y_i)^\top$ to be in the second quadrant:

$$\Pr \left\{ Z_{i-1} < -\sqrt{\mathcal{E}} \cap Z_i > -\sqrt{\mathcal{E}} \right\} = Q \left(\sqrt{\frac{\mathcal{E}}{N_0/2}} \right) Q \left(-\sqrt{\frac{\mathcal{E}}{N_0/2}} \right),$$

so,

$$P_e(D_i = 0 | D_{i-1} = 0) = 2Q \left(\sqrt{\frac{\mathcal{E}}{N_0/2}} \right) Q \left(-\sqrt{\frac{\mathcal{E}}{N_0/2}} \right).$$

Again, due to the symmetry of the problem,

$$P_e(D_i = 0 | D_{i-1} = 1) = P_e(D_i = 0 | D_{i-1} = 0) = P_e(D_i = 0),$$

and

$$P_e(D_i = 1) = P_e(D_i = 0);$$

hence

$$P_e = 2Q \left(\sqrt{\frac{\mathcal{E}}{N_0/2}} \right) Q \left(-\sqrt{\frac{\mathcal{E}}{N_0/2}} \right).$$

SOLUTION 4.

(a) In your book, it has been shown that the power spectral density is

$$S_X(f) = \frac{|\psi_{\mathcal{F}}(f)|^2}{T} \sum_k K_X[k] \exp(-j2\pi k f T),$$

where $K_X[k]$ is the autocovariance of X_i and $\psi_{\mathcal{F}}(f)$ is the Fourier transform of $\psi(t)$.

In this case, because $\{X_i\}_{i=-\infty}^{\infty}$ are i.i.d. and have zero-mean,

$$K_X[k] = \mathbb{E}[X_{i+k}X_i^*] = \mathcal{E}\mathbb{1}\{k=0\},$$

so

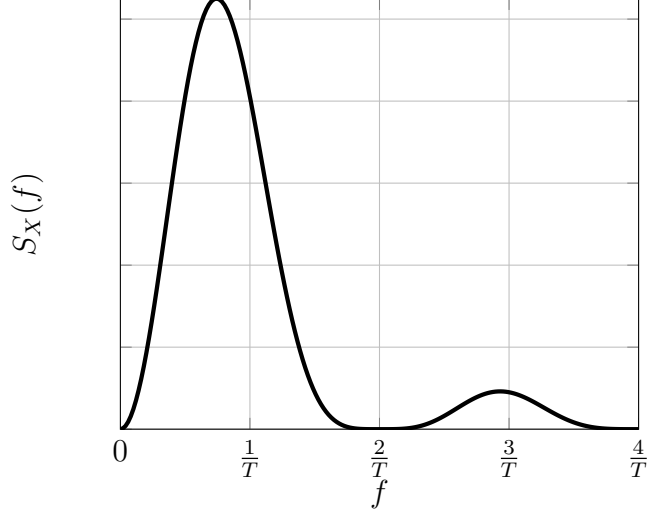
$$S_X(f) = \mathcal{E} \frac{|\psi_{\mathcal{F}}(f)|^2}{T}.$$

$$\begin{aligned} \psi_{\mathcal{F}}(f) &= \int_{-\infty}^{\infty} \psi(t) e^{-j2\pi f t} dt \\ &= \frac{1}{\sqrt{T}} \int_0^{\frac{T}{2}} e^{-j2\pi f t} dt - \frac{1}{\sqrt{T}} \int_{\frac{T}{2}}^T e^{-j2\pi f t} dt \\ &= \frac{j}{2\pi f \sqrt{T}} \left(e^{-j2\pi f \frac{T}{2}} - 1 - e^{-j2\pi f T} + e^{-j2\pi f \frac{T}{2}} \right) \\ &= \frac{j}{2\pi f \sqrt{T}} e^{-j2\pi f \frac{T}{2}} \left(2 - e^{j2\pi f \frac{T}{2}} - e^{-j2\pi f \frac{T}{2}} \right) \\ &= \frac{j}{2\pi f \sqrt{T}} e^{-j2\pi f \frac{T}{2}} (2 - 2 \cos(\pi f T)) \\ &= \frac{j}{2\pi f \sqrt{T}} e^{-j2\pi f \frac{T}{2}} 4 \sin^2 \left(\pi f \frac{T}{2} \right). \end{aligned}$$

Therefore,

$$\begin{aligned} S_X(f) &= \mathcal{E} \frac{16 \sin^4(\pi f \frac{T}{2})}{4\pi^2 f^2 T^2} \\ &= \mathcal{E} \left(\pi f \frac{T}{2} \right)^2 \text{sinc}^4 \left(f \frac{T}{2} \right). \end{aligned}$$

We first see that $S_X(0) = 0$. Furthermore, for $f > 0$, $S_X(f) = 0$ if and only if $\sin(\pi f \frac{T}{2}) = 0$. That is $f \frac{T}{2} = m \in \mathbb{Z}$ which means the spectrum is zero at frequencies $f_m = m \frac{2}{T}$, $m \in \mathbb{Z}$ (multiples of $\frac{2}{T}$). $S_X(f)$ is plotted here:



(b) Using the described precoding,

$$\begin{aligned}
K_X[k] &= \mathbb{E}[X_{i+k}X_i^*] \\
&= \mathcal{E} \mathbb{E}[(D_{i+k} + \alpha D_{i+k-1})(D_i + \alpha D_{i-1})] \\
&= \mathcal{E} (\mathbb{E}[D_{i+k}D_i] + \alpha \mathbb{E}[D_{i+k}D_{i-1}] + \alpha \mathbb{E}[D_{i+k-1}D_i] + \alpha^2 \mathbb{E}[D_{i+k-1}D_{i-1}]) \\
&= \mathcal{E} (\mathbb{1}\{k=0\} + \alpha \mathbb{1}\{k=-1\} + \alpha \mathbb{1}\{k=1\} + \alpha^2 \mathbb{1}\{k=0\}) \\
&= \begin{cases} (1 + \alpha^2)\mathcal{E} & k = 0, \\ \alpha\mathcal{E} & k = \pm 1, \\ 0 & \text{otherwise.} \end{cases}
\end{aligned}$$

Thus,

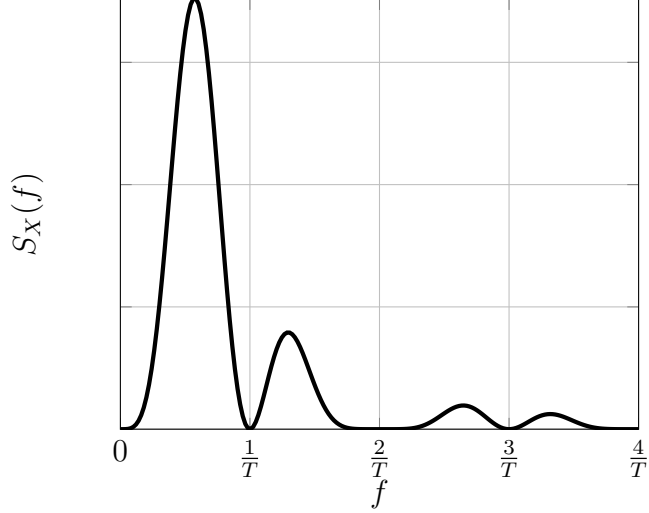
$$\begin{aligned}
\sum_k K_X[k] \exp(-j2\pi k f T) &= ((1 + \alpha^2) + \alpha \exp(j2\pi f T) + \alpha \exp(-j2\pi f T)) \mathcal{E} \\
&= (1 + \alpha^2 + 2\alpha \cos(2\pi f T)) \mathcal{E} \tag{1}
\end{aligned}$$

To create a null at $f = \frac{1}{T}$, the above should evaluate to 0 at $f = \frac{1}{T}$. That is,

$$1 + \alpha^2 + 2\alpha = (1 + \alpha)^2 = 0.$$

Thus, by choosing $\alpha = -1$ we can create the desired zero at $f = \frac{1}{T}$. In this case the power spectrum of $X(t)$ will be

$$\begin{aligned}
S_X(f) &= \mathcal{E} \frac{16 \sin^4(\pi f \frac{T}{2})}{4\pi^2 f^2 T^2} (2 - 2 \cos(2\pi f T)) \\
&= \mathcal{E} \frac{64 \sin^4(\pi f \frac{T}{2})}{4\pi^2 f^2 T^2} \sin^2(\pi f T)
\end{aligned}$$



- (c) The precoding of part (b) cannot create the desired nulls since for (1) to evaluate to 0 at $f = m\frac{1}{4T}$, we must have

$$1 + \alpha^2 + 2\alpha \cos(m\pi/2) = 0 \quad \forall m \in \mathbb{Z}.$$

However, this is impossible since for m odd we require $1 + \alpha^2 = 0$ which cannot be satisfied for any real α .

However, one can create nulls at multiples of $\frac{1}{4T}$ using a precoding of the form

$$X_i = \sqrt{\mathcal{E}}(D_i + \alpha D_{i-4}).$$

Using such a precoding, following the same steps as in (b) we can conclude that

$$K_X[k] = \begin{cases} (1 + \alpha^2)\mathcal{E} & k = 0, \\ \alpha\mathcal{E} & k = \pm 4, \\ 0 & \text{otherwise.} \end{cases}$$

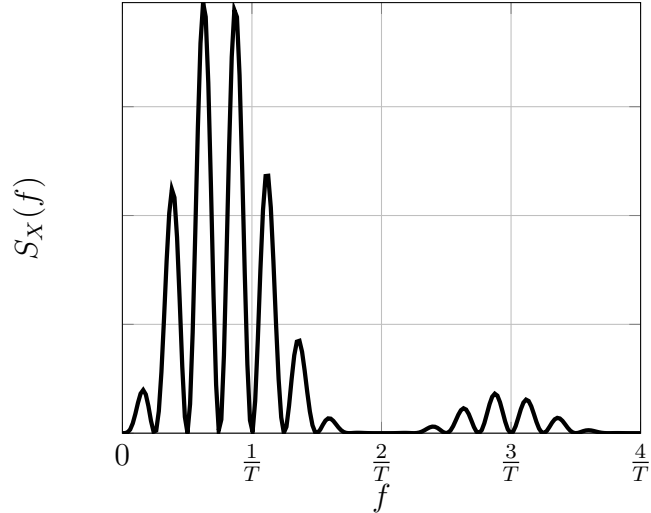
Therefore,

$$\begin{aligned} \sum_k K_X[k] \exp(-j2\pi k f T) &= ((1 + \alpha^2) + \alpha \exp(j2\pi 4 f T) + \alpha \exp(-j2\pi 4 f T)) \mathcal{E} \\ &= (1 + \alpha^2 + 2\alpha \cos(8\pi f T)) \mathcal{E} \end{aligned} \quad (2)$$

which evaluates to 0 at $f = m\frac{1}{4T}$ for $\forall m \in \mathbb{Z}$ if we choose $\alpha = -1$.

In this case, the power spectrum of $X(t)$ will similarly be

$$S_X(f) = \mathcal{E} \frac{64 \sin^4(\pi f \frac{T}{2})}{4\pi^2 f^2 T^2} \sin^2(4\pi f T)$$



SOLUTION 5. From Theorem 5.6, we know that $\{\psi(t - jT)\}_{j=-\infty}^{\infty}$ is an orthonormal set if and only if

$$\sum_{k \in \mathbb{Z}} |\psi_{\mathcal{F}}(f - \frac{k}{T})|^2 = T.$$

(a)

$$\sum_{k \in \mathbb{Z}} T \mathbb{1}_{[\frac{k}{T} - \frac{1}{2T}, \frac{k}{T} + \frac{1}{2T}]}(f) = T \Rightarrow \text{The Nyquist criterion is satisfied}$$

$\Rightarrow \psi(t)$ is orthonormal to its time-translates by multiples of T .

(b)

$$\sum_{k \in \mathbb{Z}} \frac{T}{2} \mathbb{1}_{[\frac{k-1}{T}, \frac{k+1}{T}]}(f) = T \Rightarrow \text{The Nyquist criterion is satisfied}$$

$\Rightarrow \psi(t)$ is orthonormal to its time-translates by multiples of T .

(c) Because $|\psi_{\mathcal{F}}(f)|^2$ vanishes outside $[-\frac{1}{T}, \frac{1}{T}]$, we verify whether the band-edge symmetry is fulfilled, which is the case. Hence, the Nyquist criterion is satisfied and $\psi(t)$ is orthonormal to its time-translates by multiples of T . Note: the same reasoning can be applied to (b).

(d) $\psi_{\mathcal{F}}(f)$ is a sinc function, therefore $\psi(t)$ is a box function, equal to $\frac{1}{T} \mathbb{1}_{[-\frac{T}{2}, \frac{T}{2}]}(t)$. This is orthogonal to its time-translates by multiples of T , but does not have unit norm (unless $T = 1$): $\int_{-\infty}^{\infty} |\psi(t)|^2 dt = \frac{1}{T}$.