# ÉCOLE POLYTECHNIQUE FÉDÉRALE DE LAUSANNE

School of Computer and Communication Sciences

## Handout 29

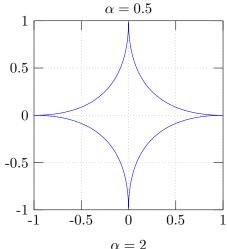
Solutions to Problem Set 12

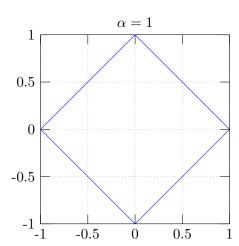
Principles of Digital Communications

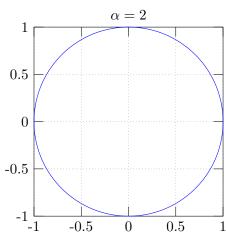
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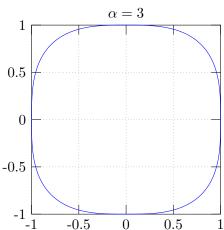
#### SOLUTION 1.

(a) (i) The plots are shown below:









- (ii) The joint density function is invariant under rotation for  $\alpha = 2$  only. For this value of  $\alpha$ , we have  $X, Y \sim \mathcal{N}\left(0, \frac{1}{2}\right)$ .
- (b) (i) We know that we can write (x, y) in polar coordinates  $(r, \theta)$ . Hence in general the joint distribution of X and Y is a function of r and  $\theta$ . Because of circular symmetry the joint distribution should not depend on  $\theta$ , which means that  $f_{X,Y}(x,y)$  can be written as a function of r. Hence if we denote this function by  $\psi$  and use the independence of X and Y, we have  $f_X(x)f_Y(y) = \psi(r)$ .
  - (ii) Taking the partial derivative with respect to x and using the chain rule for differentiation, we have  $f_X'(x)f_Y(y) = \psi'(r)\frac{\partial r}{\partial x} = \psi'(r)\frac{x}{r}$ . If we divide both sides by  $xf_X(x)f_Y(y)$  we have  $\frac{f_X'(x)}{xf_X(x)} = \frac{\psi'(r)}{r\psi(r)}$ . Proceeding similarly for y, we obtain

$$\frac{f_X'(x)}{xf_X(x)} = \frac{\psi'(r)}{r\psi(r)} = \frac{f_Y'(y)}{yf_Y(y)}.$$

- (iii)  $\frac{f_X'(x)}{xf_X(x)}$  is a function of x while  $\frac{f_Y'(y)}{yf_Y(y)}$  is a function of y. Hence the only way for the equality to hold is that both of them equal a constant. If we denote this constant by  $-\frac{1}{\sigma^2}$ , we reach the final result.
- (iv) We have  $\frac{f_X'(x)}{f_X(x)} = -\frac{x}{\sigma^2}$ . Integrating both sides we have  $\log(\frac{f_X(x)}{C}) = -\frac{x^2}{2\sigma^2}$ . Hence  $f_X(x) = C \exp(-\frac{x^2}{2\sigma^2})$ .  $f_X(x)$  is a probability density function and so should integrate to 1, which gives  $C = \frac{1}{\sqrt{2\pi\sigma^2}}$ . Hence  $f_X(x) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp(-\frac{x^2}{2\sigma^2})$  and by symmetry  $f_Y(y) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp(-\frac{y^2}{2\sigma^2})$ , which shows that X and Y are Gaussian random variables.

# SOLUTION 2.

(a) Let  $x_E(t) = x_R(t) + jx_I(t)$ . Then

$$x(t) = \sqrt{2}\Re\{x_E(t)e^{j2\pi f_c t}\}\$$
  
=  $\sqrt{2}\Re\{[x_R(t) + jx_I(t)]e^{j2\pi f_c t}\}\$   
=  $\sqrt{2}[x_R(t)\cos(2\pi f_c t) - x_I(t)\sin(2\pi f_c t)].$ 

Hence, we have

$$x_{EI}(t) = \sqrt{2}\Re\{x_E(t)\}$$

and

$$x_{EQ}(t) = \sqrt{2}\Im\{x_E(t)\}.$$

(b) Let  $x_E(t) = \alpha(t)e^{j\beta(t)}$ . Then

$$x(t) = \sqrt{2}\Re\{x_E(t)e^{j2\pi f_c t}\}$$

$$= \sqrt{2}\Re\{\alpha(t)e^{j\beta(t)}e^{j2\pi f_c t}\}$$

$$= \sqrt{2}\Re\{\alpha(t)e^{j(2\pi f_c t + \beta(t))}\}$$

$$= \sqrt{2}\alpha(t)\cos[2\pi f_c t + \beta(t)].$$

We thus have

$$x_E(t) = \alpha(t)e^{j\beta(t)} = \frac{a(t)}{\sqrt{2}}e^{j\theta(t)}.$$

(c) From (b) we see that

$$x_E(t) = \frac{A(t)}{\sqrt{2}} e^{j\varphi}.$$

This is consistent with Example 7.9 (DSB-SC) given in the text. We can also verify:

$$x(t) = \sqrt{2}\Re\{x_E(t)e^{j2\pi f_c t}\}$$

$$= \sqrt{2}\Re\left\{\frac{A(t)}{\sqrt{2}}e^{j\varphi}e^{j2\pi f_c t}\right\}$$

$$= \Re\{A(t)e^{j(2\pi f_c t + \varphi)}\}$$

$$= A(t)\cos(2\pi f_c t + \varphi).$$

## SOLUTION 3.

(a) The key observation is that while  $e^{j2\pi f_1 t}$  and  $e^{-j2\pi f_1 t}$  are two different signals if  $f_1 \neq 0$ ,  $\Re\{e^{j2\pi f_1 t}\}$  and  $\Re\{e^{-j2\pi f_1 t}\}$  are identical.

Therefore, if we fix  $f_1 \neq 0$  and choose  $a_1(t)$  and  $a_2(t)$  so that  $a_1(t)e^{j2\pi f_c t} = e^{j2\pi f_1 t}$  and  $a_2(t)e^{j2\pi f_c t} = e^{-j2\pi f_1 t}$ , we get  $a_1(t) \neq a_2(t)$  and  $\Re\left\{a_1(t)e^{j2\pi f_c t}\right\} = \Re\left\{a_2(t)e^{j2\pi f_c t}\right\}$ .

Let  $a_1(t) = e^{-j2\pi(f_c - f_1)t}$  and  $a_2(t) = e^{-j2\pi(f_c + f_1)t}$ . Then  $a_1(t) \neq a_2(t)$  and

$$\sqrt{2}\Re\left\{a_1(t)e^{\mathrm{j}2\pi f_ct}\right\} = \sqrt{2}\Re\left\{a_2(t)e^{\mathrm{j}2\pi f_ct}\right\}.$$

- (b) Let  $b(t) = a(t)e^{j2\pi f_c t}$ , which represents a translation of a(t) in the frequency domain. If  $a_{\mathcal{F}}(f) = 0$  for  $f < -f_c$ , then  $b_{\mathcal{F}}(f) = 0$  for f < 0. Because  $\Re\{b(t)\} = \frac{1}{2}\left(a(t)e^{j2\pi f_c t} + a^*(t)e^{-j2\pi f_c t}\right)$ , taking the real part has a scaling effect and adds a negative-frequency component. The negative spectrum is canceled by the  $h_{>}$  filter, and the scaling is compensated by the  $\sqrt{2}$  factors from the up-converter and down-converter. Multiplying by  $e^{-j2\pi f_c t}$  translates the spectrum back to the initial position. In conclusion, we obtain a(t).
- (c) Take any baseband signal u(t) with frequency domain support  $[-f_c \Delta, f_c + \Delta], \Delta > 0$ . The signal can be real-valued or complex-valued (for example  $u_{\mathcal{F}}(f) = \mathbb{1}_{[-f_c - \Delta, f_c + \Delta]}(f)$ , which is a sinc in time domain). After we up-convert, the support of  $u_{\mathcal{F}}(f)$  will not extend beyond  $2f_c + \Delta$ . When we chop the negative frequencies we obtain a support contained in  $[0, 2f_c + \Delta]$  and when we shift back to the left the support will be contained in  $[-f_c, f_c + \Delta]$ , which is too small to be the support of  $u_{\mathcal{F}}(f)$ .
- (d) In time domain:

$$w(t) = \sqrt{2}\Re\{a(t)e^{j2\pi f_c t}\}$$

$$\stackrel{a \in \mathbb{R}}{=} \sqrt{2}a(t)\cos(2\pi f_c t).$$

Therefore,

$$a(t) = \frac{w(t)}{\sqrt{2}\cos(2\pi f_c t)}.$$

In frequency domain: If  $a_{\mathcal{F}}(f) = 0$  for  $f < -f_c$ , we obtain a(t) as described in (b). In the following, we consider the case  $a_{\mathcal{F}}(f) \neq 0$  for  $f < -f_c$ .

We have  $w_{\mathcal{F}}(f) = \frac{1}{\sqrt{2}} \left[ a_{\mathcal{F}}(f - f_c) + a_{\mathcal{F}}(f + f_c) \right] = a_{\mathcal{F}}^+(f) + a_{\mathcal{F}}^-(f)$ , with  $a_{\mathcal{F}}^+(f) = \frac{1}{\sqrt{2}} a_{\mathcal{F}}(f - f_c)$  and  $a_{\mathcal{F}}^-(f) = \frac{1}{\sqrt{2}} a_{\mathcal{F}}(f + f_c)$ , respectively. These two components have overlapping support in some interval centered at 0. However, there is no overlap for sufficiently large frequencies. This means that for sufficiently large frequencies f we have  $w_{\mathcal{F}}(f) = \frac{1}{\sqrt{2}} a_{\mathcal{F}}^+(f)$ , which implies that from  $w_{\mathcal{F}}(f)$  we can observe the right tail of  $a_{\mathcal{F}}^+(f)$  and use that information to remove the right tail of  $a_{\mathcal{F}}^-(f)$  from  $w_{\mathcal{F}}(f)$  (the right tails of  $a_{\mathcal{F}}^+(f)$  and  $a_{\mathcal{F}}^-(f)$  are the same because a(t) is real). Hence, from  $w_{\mathcal{F}}(f)$  we can read more of the right tail of  $a_{\mathcal{F}}^+(f)$ . The procedure can be repeated until we get to see  $a_{\mathcal{F}}^+(f)$  for all frequencies above  $f_c$ . At this point, using  $a_{\mathcal{F}}(f) = a_{\mathcal{F}}^+(f + f_c)\sqrt{2}$  and the fact that a(t) is real-valued, we have  $a_{\mathcal{F}}(f)$  for the positive frequencies, hence for all frequencies.

SOLUTION 4.

$$x(t)\sqrt{2}\cos(2\pi f_c t) = x(t) \left[ \frac{e^{j2\pi f_c t} + e^{-j2\pi f_c t}}{\sqrt{2}} \right]$$

$$= \sqrt{2}\Re\{x_E(t)e^{j2\pi f_c t}\} \left[ \frac{e^{j2\pi f_c t} + e^{-j2\pi f_c t}}{\sqrt{2}} \right]$$

$$= \left[ \frac{x_E(t)e^{j2\pi f_c t} + x_E^*(t)e^{-j2\pi f_c t}}{\sqrt{2}} \right] \left[ \frac{e^{j2\pi f_c t} + e^{-j2\pi f_c t}}{\sqrt{2}} \right]$$

$$= \frac{x_E(t)e^{j4\pi f_c t} + x_E(t) + x_E^*(t) + x_E^*(t)e^{-j4\pi f_c t}}{\sqrt{2}}.$$

At the lowpass filter output we have

$$\frac{x_E(t) + x_E^*(t)}{2} = \Re\{x_E(t)\}.$$

The calculation for the other path is similar.

SOLUTION 5.

- (a) Notice that the sinusoids of w(t) have a period of  $T_c=4$  ms units of time, which implies that  $f_c=\frac{1}{T_c}=\frac{1}{4\,\mathrm{ms}}=250\,\mathrm{Hz}.$
- (b) Notice that the phase of the sinusoidal signal changes every  $T_s = 4$  ms. (Here we have  $T_s = T_c$ , but in general it is not the case. In practice we usually have  $T_s \gg T_c$ . See the note at the end.)

The expression of w(t) as a function of t is:

$$w(t) = \begin{cases} 4\cos(2\pi f_c t - \frac{\pi}{2}) & t \in ]0, T_s[\\ 4\cos(2\pi f_c t) & t \in ]T_s, 2T_s[\\ 4\cos(2\pi f_c t + \pi) & t \in ]2T_s, 3T_s[\\ 4\cos(2\pi f_c t + \frac{\pi}{2}) & t \in ]3T_s, 4T_s[ \end{cases} = \begin{cases} \Re\left\{4e^{\mathrm{j}(2\pi f_c t - \frac{\pi}{2})}\right\} & t \in ]0, T_s[\\ \Re\left\{4e^{\mathrm{j}(2\pi f_c t)}\right\} & t \in ]T_s, 2T_s[\\ \Re\left\{4e^{\mathrm{j}(2\pi f_c t + \pi)}\right\} & t \in ]2T_s, 3T_s[\\ \Re\left\{4e^{\mathrm{j}(2\pi f_c t + \pi)}\right\} & t \in ]2T_s, 3T_s[\\ \Re\left\{4e^{\mathrm{j}(2\pi f_c t + \pi)}\right\} & t \in ]3T_s, 4T_s[ \end{cases}$$
$$= \begin{cases} \Re\left\{-4\mathrm{j}e^{\mathrm{j}2\pi f_c t}\right\} & t \in ]0, T_s[\\ \Re\left\{4e^{\mathrm{j}(2\pi f_c t + \pi)}\right\} & t \in ]3T_s, 2T_s[\\ \Re\left\{4e^{\mathrm{j}(2\pi f_c t + \pi)}\right\} & t \in ]3T_s, 3T_s[\\ \Re\left\{4e^{\mathrm{j}(2\pi f_c t + \pi)}\right\} & t \in ]3T_s, 3T_s[\\ \Re\left\{4e^{\mathrm{j}(2\pi f_c t + \pi)}\right\} & t \in ]3T_s, 3T_s[\\ \Re\left\{4e^{\mathrm{j}(2\pi f_c t + \pi)}\right\} & t \in ]3T_s, 4T_s[ \end{cases}$$

where

$$\begin{split} w_E(t) &= -\frac{4\mathrm{j}}{\sqrt{2}}\mathbbm{1}\{t\in]0, T_s[\} + \frac{4}{\sqrt{2}}\mathbbm{1}\{t\in]T_s, 2T_s[\} \\ &- \frac{4}{\sqrt{2}}\mathbbm{1}\{t\in]2T_s, 3T_s[\} + \frac{4\mathrm{j}}{\sqrt{2}}\mathbbm{1}\{t\in]3T_s, 4T_s[\} \\ &= -\mathrm{j}\sqrt{8T_s}\frac{1}{\sqrt{T_s}}\mathbbm{1}\{t\in]0, T_s[\} + \sqrt{8T_s}\frac{1}{\sqrt{T_s}}\mathbbm{1}\{t\in]T_s, 2T_s[\} \\ &- \sqrt{8T_s}\frac{1}{\sqrt{T_s}}\mathbbm{1}\{t\in]2T_s, 3T_s[\} + \mathrm{j}\sqrt{8T_s}\frac{1}{\sqrt{T_s}}\mathbbm{1}\{t\in]3T_s, 4T_s[\}. \end{split}$$

If we define  $\psi(t) = \frac{1}{\sqrt{T_s}} \mathbb{1}\{t \in ]0, T_s[\}, c_0 = -j\sqrt{8T_s}, c_1 = \sqrt{8T_s}, c_2 = -\sqrt{8T_s} \text{ and } c_3 = j\sqrt{8T_s}, \text{ we get}$ 

$$w_E(t) = \sum_{i=0}^{3} c_i \psi(t - iT_s).$$
 (1)

Therefore, the pulse used in the waveform former is  $\psi(t) = \frac{1}{\sqrt{T_s}} \mathbb{1}\{t \in ]0, T_s[\}$ , and the waveform former output signal is given by (1). The orthonormal basis that is used is  $\{\psi(t-iT_s)\}_{i=0}^3$ .

- (c) The symbol sequence is  $\{c_0, c_1, c_2, c_3\} = \{-j\sqrt{\mathcal{E}_s}, \sqrt{\mathcal{E}_s}, -\sqrt{\mathcal{E}_s}, j\sqrt{\mathcal{E}_s}\}$ , where  $\mathcal{E}_s = 8T_s$ . We can see that the symbol alphabet is  $\{\sqrt{\mathcal{E}_s}, j\sqrt{\mathcal{E}_s}, -\sqrt{\mathcal{E}_s}, -j\sqrt{\mathcal{E}_s}\}$ .
- (d) We have:
  - The output sequence of the encoder is the symbol sequence, which is

$$\{c_0, c_1, c_2, c_3\} = \left\{-j\sqrt{\mathcal{E}_s}, \sqrt{\mathcal{E}_s}, -\sqrt{\mathcal{E}_s}, j\sqrt{\mathcal{E}_s}\right\}.$$

- The symbol alphabet contains 4 symbols. This means that each symbol represents two bits. Since the symbol rate is  $f_s = \frac{1}{T_s} = 250$  symbols/s, the bit rate is  $2 \times 250 = 500$  bits/s.
- The input/output mapping can be obtained by assigning two bits for each symbol in the symbol alphabet. Keeping in mind that it is better to minimize the number of bit-differences between close symbols, we obtain the following input/output mapping (which is not unique, i.e., we can obtain other mappings that satisfy the mentioned criterion):  $\sqrt{\mathcal{E}_s} \longleftrightarrow 00$ ,  $j\sqrt{\mathcal{E}_s} \longleftrightarrow 01$ ,  $-\sqrt{\mathcal{E}_s} \longleftrightarrow 11$  and  $-j\sqrt{\mathcal{E}_s} \longleftrightarrow 10$ .
- Assuming that the above input/output mapping was used, we can obtain the input sequence of the encoder: 10001101.

Note that in this example, we have  $T_s = T_c$ , so  $f_c = f_s$ . This is very unusual. In practice we almost always have  $f_c \gg f_s$ , especially if we are using electromagnetic waves.