

EE-120: Homework 2

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We can say the signal \hat{x} is equal to the following.

$$\hat{x}(t) = x(t - \frac{1}{2}) - x(t + \frac{1}{2}) \quad (0.1)$$

Where x is the square wave.

By linearity: We know that given the two signals $x(t - \frac{1}{2})$ and $x(t + \frac{1}{2})$ with FS coefficients a'_k, b'_k then \hat{x} will have FS coefficients equal to $a'_k - b'_k$

By time shift on the signal $x(t - \frac{1}{2})$ we know that $a'_k = a_k e^{-jk\frac{\pi}{2}}$ Where a_k are the FS coefficients of the square wave.

Similarly, by time shift on the signal $x(t + \frac{1}{2})$ we know that $b'_k = a_k e^{jk\frac{\pi}{2}}$

This implies that the fourier series coefficients for the signal \hat{x} are $a'_k - b'_k$:

$$= a_k e^{-jk\frac{\pi}{2}} - a_k e^{jk\frac{\pi}{2}} \quad (0.2)$$

$$= -a_k \left(-e^{-jk\frac{\pi}{2}} + e^{jk\frac{\pi}{2}} \right) \quad (0.3)$$

$$= -a_k \left(e^{jk\frac{\pi}{2}} + -e^{-jk\frac{\pi}{2}} \right) \quad (0.4)$$

$$= -2j a_k \sin(k\frac{\pi}{2}) \quad (0.5)$$

$$= \frac{-2j}{k\pi} \sin^2(\frac{\pi k}{2}) \quad (0.6)$$

$$= \begin{cases} 0 : k \text{ is even} \\ \frac{-2j}{k\pi} : k \text{ is odd} \end{cases} \quad (0.7)$$

- a: Let $x: \mathbb{R} \rightarrow \mathbb{R}$ have FS coefficients a_k and let $x(t) = x(-t) \forall t$
Because x is real valued then $x^* = x$ which tells us $a_k = a_{-k}^*$ and Because x is even symmetric, $a_k = a_{-k}$. Because $a_{-k} = a_{-k}^* \forall k$
This implies the fourier series coefficients are purely real.

- b: Lemma: $x = a + bi \in \mathbb{C} - x = x^* \rightarrow a = 0$
By definition we define the $x^* = a - bi$

$$x^* = a - bi = -a - bi = -x \quad (0.8)$$

Which only holds true for $a = 0$

Let x be defined as above with FS coefficients and let $x(t) = -x(-t)$.

Combining the time shift and linearity properties of FS coefficients, $a_k = -a_{-k}$ and because x is real valued, then $a_k = a_{-k}^*$

This implies $a_{-k} = a_{-k}^* : \forall k$. This implies the Fourier series coefficients are purely imaginary.

- c: Let x be defined as above with FS coefficients and let $x(t) = -x(t + \frac{T}{2})$
By the linearity and time-shift properties of FS coefficients, this implies that $a_k = -a_k e^{-jk\omega_0 t_0}$, with $\omega_0 = \frac{2\pi}{T}$ and $t_0 = -\frac{T}{2}$
This implies $a_k = -a_k e^{-jk\pi}$
Let $k = 2a : a \in \mathbb{Z}$
 $a_k = a_{2a} = -a_{2a} e^{-j2a\pi} = -a_{2a} * 1$ Which is only true for $a_{2a} = 0$

- d: Consider $t \in [0, \frac{T}{2}]$ and let $x(t) = 1$
 $|t| < T_1 \rightarrow |-t| < T_1 \rightarrow x(-t) = 1$
Now, if we let $x(t) = 0$
 $T_1 < |t| \leq \frac{T}{2} \rightarrow T_1 < |-t| \leq \frac{T}{2} \rightarrow x(-t) = 0$. Because this is true over all t in a given period, Then the square wave is even symmetric, and as we can see, the Fourier series coefficients a_k for the square wave are purely real.

Now, consider the signal $\hat{x}(t) = x(t - \frac{1}{2}) - x(t + \frac{1}{2})$, where x is the square wave defined earlier.

$$-\hat{x}(-t) = -x(-t - \frac{1}{2}) + x(-t + \frac{1}{2}) \quad (0.9)$$

$$= -x(-(t + \frac{1}{2})) + x(-(t - \frac{1}{2})) \quad (0.10)$$

$$= x(t + \frac{1}{2}) + x(t - \frac{1}{2}) \quad (0.11)$$

$$= \hat{x}(t) \quad (0.12)$$

Which means \hat{x} is odd symmetric which we can see from the FS coefficients which are

all imaginary. Setting $T = 2$, we can see that property 3 holds.

$$-x(t+1) = -x\left((t+1) - \frac{1}{2}\right) + x\left((t+1) + \frac{1}{2}\right) \quad (0.13)$$

$$= -x\left(t + \frac{1}{2}\right) + x\left(t + \frac{3}{2}\right) \quad (0.14)$$

$$= -x\left(t + \frac{1}{2}\right) + x\left(t - \frac{1}{2}\right) \quad (0.15)$$

$$= \hat{x}(t) \quad (0.16)$$

And as we can see, the following signal has $a_k = 0$ for even k .

- a: Let $W := [\Phi_0 \quad \Phi_1 \quad \dots \quad \Phi_{N-1}]$

$$\Phi_k = \begin{bmatrix} 1 \\ e^{-\frac{jk2\pi}{N}} \\ \vdots \\ e^{-\frac{jk2\pi}{N}(N-1)} \end{bmatrix}$$

- b:

Done in iPython Notebook. The Fourier series coefficients are $a_0 = a_1 = a_2 = \frac{1}{3}$

- c:

Given the analysis equation in class we can define $W^{-1} = \frac{1}{N} [\Phi_0^{-1} \quad \Phi_1^{-1} \quad \dots \quad \Phi_{N-1}^{-1}]$

$$\Phi_k^{-1} = \begin{bmatrix} 1 \\ e^{\frac{jk2\pi}{N}} \\ \vdots \\ e^{\frac{jk2\pi}{N}(N-1)} \end{bmatrix}$$

- d: Performed in iPython Notebook.

- a: Given $X(\omega)$ equal to the following:

$$X(\omega) = \int_{-\infty}^{\infty} e^{-t^2} e^{-j\omega t} dt \quad (0.17)$$

Then $\frac{\partial X(\omega)}{\partial \omega}$ is equal to the following:

$$\frac{\partial X(\omega)}{\partial \omega} = \int_{-\infty}^{\infty} -j t e^{-t^2} e^{-j\omega t} dt \quad (0.18)$$

$$= \frac{j}{2} \left(e^{-j\omega t} e^{-t^2} \Big|_{-\infty}^{\infty} + \int_{-\infty}^{\infty} e^{-t^2} j\omega e^{-j\omega t} dt \right) \quad (0.19)$$

$$= \frac{j^2}{2} \left(\int_{-\infty}^{\infty} e^{-t^2} \omega e^{-j\omega t} dt \right) \quad (0.20)$$

$$= \frac{\omega}{2} \left(\int_{-\infty}^{\infty} e^{-t^2} e^{-j\omega t} dt \right) \quad (0.21)$$

$$= -\frac{\omega}{2} X(\omega) \quad (0.22)$$

- b: To find α, β given $X(\omega) = \alpha e^{-\frac{\omega^2}{\beta}}$ we need to find the unique solutions to the following system of equations.

$$X(0) = \sqrt{\pi} \quad (0.23)$$

$$\frac{dX(\omega)}{d\omega} = -\frac{\omega}{2} X(\omega) \quad (0.24)$$

Starting from eq. 0.23

$$X(0) = \alpha e^{\frac{0}{\beta}} = \alpha = \sqrt{\pi} \quad (0.25)$$

Now with eq. 0.24

$$\frac{dX(\omega)}{d\omega} = -\frac{\omega}{2} X(\omega) \quad (0.26)$$

$$-\frac{2\alpha\omega}{\beta} e^{-\frac{\omega^2}{\beta}} = -\frac{\omega}{2} \alpha e^{-\frac{\omega^2}{\beta}} \quad (0.27)$$

$$\frac{2}{\beta} = \frac{1}{2} \quad (0.28)$$

$$\beta = 4 \quad (0.29)$$

Which means $\alpha = \sqrt{\pi}, \beta = 4$

Given the following modulation scheme:

$$r(t) = s(t) \cos(\omega_0 t + \phi) \quad (0.30)$$

$$= s(t) \left(\frac{e^{-j\omega_0 t + \phi} + e^{j(\omega_0 t + \phi)}}{2} \right) \quad (0.31)$$

$$\rightarrow R(\omega) = \frac{1}{2} e^{-j\phi} S(\omega + \omega_0) + \frac{1}{2} e^{j\phi} S(\omega - \omega_0) \quad (0.32)$$

$$g(t) = r(t) \cos(\omega_0 t) \quad (0.33)$$

$$\rightarrow G(\omega) = \frac{1}{2} R(\omega - \omega_0) + \frac{1}{2} R(\omega + \omega_0) \quad (0.34)$$

$$= \frac{1}{2} \left(\frac{1}{2} (e^{-j\phi} S(\omega) + e^{j\phi} S(\omega) + e^{-j\phi} S(\omega + 2\omega_0) + e^{j\phi} S(\omega - 2\omega_0)) \right) \quad (0.35)$$

$$= \frac{1}{2} \cos(\phi) S(\omega) + \frac{1}{4} (e^{-j\phi} S(\omega + 2\omega_0) + e^{j\phi} S(\omega - 2\omega_0)) \quad (0.36)$$

After $g(t)$ is run through an ideal lowpass filter, with passband gain of 2 $Y(\omega)$ is equal to the following:

$$= \cos(\phi) S(\omega) \quad (0.37)$$

$$\rightarrow y(t) = \cos(\phi) s(t) \quad (0.38)$$

In the case $\phi = \frac{\pi}{2}$, then the receiver does not receive $s(t)$ because the signal was attenuated completely. The signal is completely recovered when ϕ is an integer multiple of 2π

HW3 question 3

September 26, 2019

```
[1]: import numpy as np
```

```
[2]: phi_0 = np.exp(np.array([0,
                               0,
                               0]))
phi_1 = np.exp(np.array([0,
                          0+2 * np.pi / 3j,
                          0+4 * np.pi / 3j]))
phi_2 = np.exp(np.array([0,
                          0+4 * np.pi / 3j,
                          0+8 * np.pi / 3j]))
```

```
[3]: W = np.array([phi_0, phi_1, phi_2]).T
impulse_train = np.array([1,0,0])
```

```
[4]: fs_coefficients = np.linalg.solve(W, impulse_train)
```

```
[5]: fs_coefficients
```

```
[5]: array([0.33333333+1.94289029e-16j, 0.33333333-8.32667268e-17j,
           0.33333333-1.11022302e-16j])
```

```
[6]: W @ fs_coefficients
```

```
[6]: array([ 1.00000000e+00+0.00000000e+00j, -1.66533454e-16+2.22044605e-16j,
           0.00000000e+00+5.55111512e-17j])
```

```
[7]: phi_0_inv = np.exp(np.array([0,
                                    0,
                                    0]))
phi_1_inv = np.exp(np.array([0,
                              0-2 * np.pi / 3j,
                              0-4 * np.pi / 3j]))
phi_2_inv = np.exp(np.array([0,
                              0-4 * np.pi / 3j,
                              0-8 * np.pi / 3j]))
```

```
[8]: W_inv = np.array([phi_0_inv, phi_1_inv, phi_2_inv]).T / 3
```

As we can see, we can say with a more than reasonable degree of certainty that this is the identity, confirming that W_inv is an inverse to the W matrix defined.

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[9]: np.allclose(W @ W_inv, np.eye(3), 1e-15)
```

```
[9]: True
```

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[ ]:
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