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Chapter 1

Fourier Series

The frequency spectrum is a complex-valued function of the frequency variable, and thus it is usually specified in terms of an amplitude spectrum and a phase spectrum [1]. The complex exponential form is given by:

$$x(t) = \sum_{-\infty}^{\infty} X_n e^{jn\omega_0 t}$$

$$X_n = \int_{-T/2}^{T/2} x(t) e^{-jn\omega_0 t} dt$$
(1.1)

In the next exercises we will also be using the first and/or the second derivative of the previous expressions (??), and therefore we write them here explicitly:

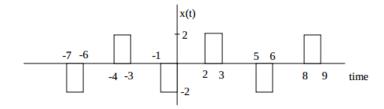
$$\dot{x}(t) = \sum_{-\infty}^{\infty} jn\omega_0 X_n e^{jn\omega_0 t}$$

$$jn\omega_0 X_n = \int_{-T/2}^{T/2} \dot{x}(t) e^{-jn\omega_0 t} dt$$
(1.2)

$$\ddot{x}(t) = \sum_{-\infty}^{\infty} -n^2 \omega_0^2 X_n e^{jn\omega_0 t}$$

$$-n^2 \omega_0^2 X_n = \int_{-T/2}^{T/2} \dot{x}(t) e^{-jn\omega_0 t} dt$$
(1.3)

For the following signal:



- a) Find the Fourier series.
- b) Plot the spectra versus frequency, $\omega = n\omega_0$.

Solution

The period of the shown signal is T=6 and therefore $\omega_0=\frac{2\pi}{T}=\frac{\pi}{3}.$

Taking the derivative of x(t) we get:

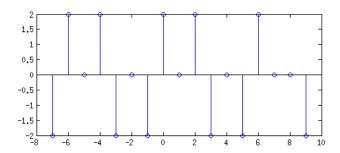


Figure 1.1: Derivative \dot{x}

The range t = [-3,3] contains one complete period of the signal. Usign (1.2) we have:

$$\dot{x}(t) = 2(-\delta(t+1) + \delta(t) + \delta(t-2) - \delta(t-3))$$

The Fourier coefficients are obtained with:

$$jn\omega_{0}X_{n} = \frac{2}{6} \int_{-3}^{3} (-\delta(t+1) + \delta(t) + \delta(t-2) - \delta(t-3))e^{-jn\omega_{0}t} dt$$

$$= \frac{1}{3} (-e^{jn\omega_{0}} + 1 + e^{-2jn\omega_{0}} - e^{-3jn\omega_{0}})$$

$$= \frac{1}{3} [e^{\frac{jn\omega_{0}}{2}} (e^{\frac{-jn\omega_{0}}{2}} - e^{\frac{jn\omega_{0}}{2}}) - e^{\frac{-5jn\omega_{0}}{2}} (e^{\frac{-jn\omega_{0}}{2}} - e^{\frac{jn\omega_{0}}{2}})]$$

$$= \frac{1}{3} [(e^{\frac{-jn\omega_{0}}{2}} - e^{\frac{jn\omega_{0}}{2}})(e^{-jn\omega_{0}} (e^{\frac{3jn\omega_{0}}{2}} - e^{\frac{-3jn\omega_{0}}{2}}))]$$

$$= \frac{4j}{3} [\sin \frac{n\omega_{0}}{2} \sin \frac{3n\omega_{0}}{2} e^{-jn\omega_{0}}]$$

$$X_{n} = \frac{-4j}{n\pi} [\sin(n\frac{\pi}{6}) \sin(n\frac{\pi}{2}) e^{-jn\frac{\pi}{3}}]$$

Next we use Matlab to plot the magnitude and phase of the spectra using the script given in [2]

Listing 1.1: Calculate and plot magnitude and phase of Xn

```
1    n=1:15;
2    Xn=-4*j./n/pi.*sin(pi*n/6).*sin(n*pi/2).*exp(-j*n*pi/3);
3    n=-15:-1;
4    X_n=-4*j./n/pi.*sin(pi*n/6).*sin(n*pi/2).*exp(-j*n*pi/3);
5    Xn=[X_n 0 Xn];
6    n=-15:15;
```

```
7    subplot(211), stem(n, abs(Xn));
8    title('|X_n|')
9    subplot(212), stem(n, angle(Xn))
10    title('angle(X_n) in rad')
```

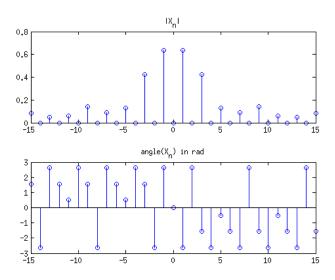


Figure 1.2: Magnitude and Angle X_n

We then plot the approximation of the function using its Fourier coefficients [2].

Listing 1.2: Approximation of x(t) with Fourier coefficients

```
function [x,t] = fapprox(N,T)

t = -1.5*T:T/1000:1.5*T;

w0 = 2*pi/T;

X0 = 0;

x = X0*ones(1,length(t)); % dc component

for n=1:N,

Xn = -4*j/n/pi*sin(pi*n/6)*sin(n*pi/2)*exp(-j*n*pi/3);

X_n = conj(Xn);

x = x + Xn*exp(j*n*w0*t) + X_n*exp(-j*n*w0*t);

end

end
```

We do this for N=5 and N=50:

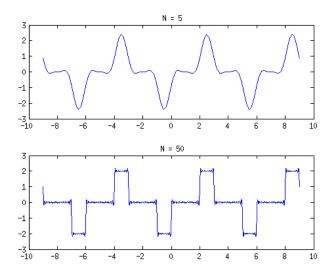
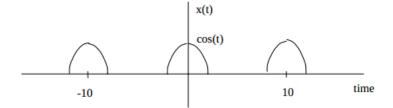


Figure 1.3: Approximation of x(t) by Xn

As we can see, the larger the N the more close to the original function we get. However, as a consequence of the Gibbs effect, we can't say that it'll be equal.

Problem 2

Repeat problem 1 for the following signal:



Solution

The period of the shown signal is T=10 and therefore $\omega_0=\frac{2\pi}{T}=\frac{\pi}{5}$.

If we take the first and second derivative of $\boldsymbol{x}(t)$ we get:

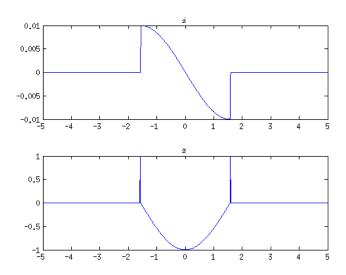


Figure 1.4: Derivative \dot{x}

The range t = [-5,5] contains one complete period of the signal. Applying (1.2) we have:

$$\ddot{x}(t) = -x(t) + \delta(t + \pi/2) + \delta(t - \pi/2)$$

$$\sum_{-\infty}^{\infty} -n^2 \omega_0^2 X_n e^{jn\omega_0 t} = \sum_{-\infty}^{\infty} X_n e^{jn\omega_0 t} + \delta(t + \pi/2) + \delta(t - \pi/2)$$

$$\sum_{-\infty}^{\infty} (1 - n^2 \omega_0^2) X_n e^{jn\omega_0 t} = \delta(t + \pi/2) + \delta(t - \pi/2)$$

Whe can now obtain X_n with:

$$(1 - n^2 \omega_0^2) X_n = \frac{1}{T} \int_{-5}^5 \delta(t + \pi/2) + \delta(t - \pi/2) e^{-jn\omega_0 t} dt$$
$$= \frac{1}{T} (e^{jn\frac{\omega_0}{2}} + e^{-jn\frac{\omega_0}{2}})$$
$$X_n = \frac{1}{5(1 - \frac{n^2 \pi^2}{25})} \cos(\frac{n\pi^2}{10})$$

Next we use Matlab to plot the magnitude and phase of the spectra using the script given in [2]

Listing 1.3: Calculate and plot magnitude and phase of Xn

```
1  n=-10:10;
2  Xn=cos(pi/2*n*w0)/5./(1-(n*w0).^2);
3  subplot(121), stem(n, abs(Xn))
4  title('|X_n|')
5  subplot(122), stem(n, angle(Xn))
6  title('angle(X_n) in rad')
```

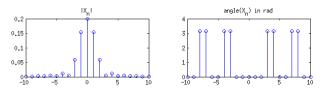


Figure 1.5: Magnitude and Angle X_n

We then plot the approximation of the function using its Fourier coefficients [2].

Listing 1.4: Approximation of x(t) with Fourier coefficients

```
function [x,t] = fapprox2(N,T)
w0 = 2*pi/T;
t = -1.5*T:T/1000:1.5*T;
c0 = 1/5;
x = c0*ones(1,length(t)); % dc component
for n=1:N,
cn = cos(pi/2*n*w0)/5/(1-(n*w0)^2);
c_n = cn;
x = x + cn*exp(j*n*w0*t) + c_n*exp(-j*n*w0*t);
end
plot(t,x)
title([' N = ',num2str(N)])
```



Figure 1.6: Approximation of x(t) by Xn

Compute the Fourier series for the following signals:

1.
$$x(t) = 2 + 4\cos(50t + \pi/2) + 12\cos(100t - \pi/3)$$

Solution

Lets recall from [1] that:

$$x(t) = a_0 + \sum_{k=1}^{\infty} A_k \cos(k\omega_0 t + \theta_k) - \infty < t < \infty$$

And the equivalent coefficients between the trigonometric series and the exponential form are:

$$X_0 = a_0$$

$$|X_n| = \frac{1}{2} A_k , k = 1, 2, ..$$

$$\angle X_n = \theta_k , k = 1, 2, ..$$
 (1.4)

From (1.4) we can immediately calculate the \mathcal{X}_n coefficients:

Listing 1.5: Plot Magnitude and Angle of X_n

```
1  fe = [-100 -50 0 50 100];
2  Ae = [6 2 2 2 6];
3  pe = [pi/3 -pi/2 0 pi/2 -pi/3];
4  
5  figure(1);
6  subplot(2,1,1), stem(fe,Ae);
7  grid on;
```

```
8 xlabel('Frec [rad/s]');
9 ylabel('Amplitude');
10 title('|X_n|', 'fontweight', 'bold');
11
12 subplot(2,1,2), stem(fe,pe);
13 grid on;
14 xlabel('Frec [rad/s]');
15 ylabel('Phase');
16 title('$\angle x$','interpreter','latex')
```

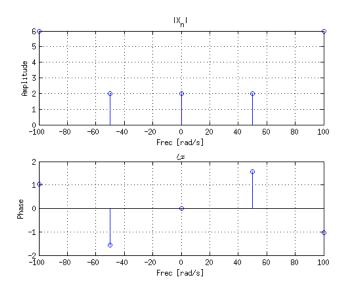


Figure 1.7: Magnitude and Angle of X_n

2.
$$x(t) = 4\cos(2\pi(1000)t)\cos(2\pi(750000)t)$$

Solution

$$\begin{split} x(t) &= 4 \left(\frac{e^{j2\pi(1000)t} + e^{-j2\pi(1000)t}}{2} \right) \left(\frac{e^{j2\pi(750000)t} + e^{-j2\pi(750000)t}}{2} \right) \\ &= e^{j2\pi(751000)t} + e^{j2\pi(-749000)t} + e^{j2\pi(749000)t} + e^{j2\pi(-751000)t} \end{split}$$

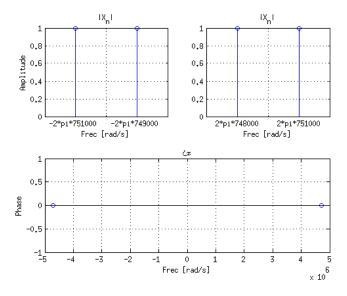
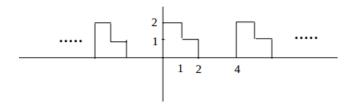


Figure 1.8: Magnitude and Angle of \mathcal{X}_n

3. The function:



Solution

The period of the shown signal is T=4 and therefore $\omega_0=\frac{2\pi}{T}=\frac{\pi}{2}.$

Taking the derivative of the function we get:

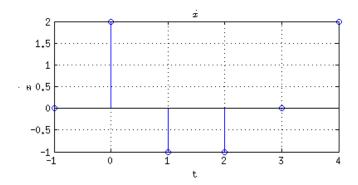


Figure 1.9: Derivative \dot{x}

In the range [-1, 3] we have:

$$\dot{x}(t) = 2\delta(t) - \delta(t-1) - \delta(t-2)$$

Applying (1.2) we have:

$$jn\omega_0 X_n = \frac{1}{T} \int_{-T/2}^{T/2} [2\delta(t) - \delta(t-1) - \delta(t-2)] e^{-jn\omega_0 t} dt$$
$$= \frac{1}{4} [2 - e^{-jn\frac{\pi}{2}} - e^{nj\pi}]$$

We can use the following properties:

$$e^{-jn\pi} = (e^{-j\pi})^n = [\cos(\pi) - j\sin(\pi)]^n = (-1)^n$$
$$e^{-jn\frac{\pi}{2}} = (e^{-j\frac{\pi}{2}})^n = [\cos(\frac{\pi}{2}) - j\sin(\frac{\pi}{2})]^n = (-j)^n$$

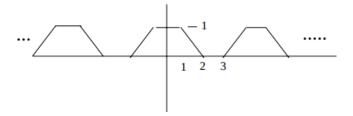
And reduce the equation to:

$$X_n = \frac{1}{2jn\pi} [2 - (-1)^n - (-j)^n]$$

where

$$X_0 = \frac{1}{4} \int_{-1}^{3} x(t) dt = \frac{3}{4}$$

4. The function:



Solution

The period of the shown signal is T=5 and therefore $\omega_0=\frac{2\pi}{T}=\frac{2\pi}{5}.$

Taking the derivative of the function we get:

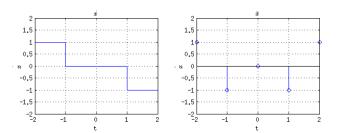


Figure 1.10: First and Second Derivatives \dot{x} \ddot{x}

In the range [-2, 2] we have:

$$\ddot{x}(t) = \delta(t+2) - \delta(t+1) - \delta(t-1) + \delta(t-2)$$

Applying (1.3) we have:

$$-n^{2}\omega_{0}^{2}X_{n} = \frac{1}{T} \int_{-T/2}^{T/2} [\delta(t+2) - \delta(t+1) - \delta(t-1) + \delta(t-2)]e^{-jn\omega_{0}t} dt$$

$$= \frac{1}{5} [(e^{2jn\omega_{0}} + e^{-2jn\omega_{0}}) - (e^{jn\omega_{0}} + e^{-jn\omega_{0}})]$$

$$X_{n} = \frac{2}{5n^{2}\omega_{0}^{2}} [\cos(n\omega_{0}) - \cos(2n\omega_{0})]$$

We can calculate the dc component by finding the area of the trapezoid:

$$X_0 = \frac{1}{T} \int_{-T/2}^{T/2} x(t) dt$$
$$= \frac{1}{5} \frac{B \cdot b}{2} h = \frac{3}{5}$$

Problem 4

For the signals given in Problem 3c) and 3d), use Matlab to plot the truncated Fourier series for N = 3, N = 10 and N = 40. (Use subplot to save paper).

Solution

1. For problem 3c:

Listing 1.6: Approximation of x(t) with Fourier coefficients

```
function [x,t] = fapprox3(N,T)
t = -1.5*T:T/1000:1.5*T;
w0 = 2*pi/T;
X0 = 3/4;

n_p = [1:N];
x_n = [-N:-1];
Xn = (2 - (-1).^n_p - (-j).^n_p) ./ (2*j*pi.*n_p);
X_n = (2 - (-1).^n_n - (-j).^n_n) ./ (2*j*pi.*n_n);
Xn = [X_n X0 Xn];
x_n = [n_n 0 n_p];
```

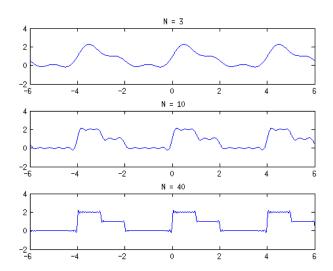


Figure 1.11: Approximation of x(t) by Xn for N=[3,10,30]

2. For problem 3d:

Listing 1.7: Approximation of x(t) with Fourier coefficients

```
function [x,t] = fapprox4(N,T)
        t = -1.5 *T:T/1000:1.5 *T;
        w0 = 2 * pi/T;
        X0 = 3/\hat{5};
4
        n_p = [1:N];
        n_n = [-N:-1];
        Xn = (\cos(n_p * w0) - \cos(2*n_p * w0)) * 2 ./ (5*w0^2*n_p.^2);
        X_n = (\cos(n_n * w0) - \cos(2*n_n * w0)) * 2 ./ (5*w0^2*n_n^2);
10
11
        Xn = [X_n X0 Xn];
12
13
        n = [n_n \ 0 \ n_p];
14
15
        x = Xn*exp(j*w0*n'*t);
16
        x = real(x);
17 end
```

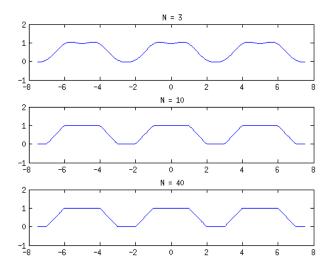
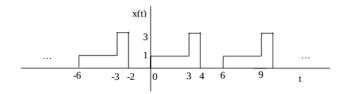


Figure 1.12: Approximation of x(t) by Xn for N=[3,10,30]

Find the Fourier series for the following signal.



Also, sketch the approximation if a large number of terms are kept in the series (say N=30).

Solution

The period of the shown signal is T=6 and therefore $\omega_0=\frac{2\pi}{T}=\frac{\pi}{3}.$

Taking the derivative of the function we get:

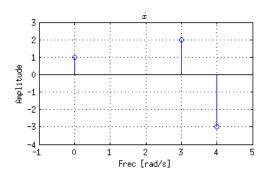


Figure 1.13: Derivative \dot{x}

In the range [-1, 5] we have:

$$\dot{x}(t) = \delta(t) + 2\delta(t-3) - 3\delta(t-4)$$

Applying (1.2) we have:

$$jn\omega_0 X_n = \frac{1}{T} \int_{-T/2}^{T/2} [\delta(t) + 2\delta(t-3) - 3\delta(t-4)] e^{-jn\omega_0 t} dt$$

$$= \frac{1}{6} [1 + 2e^{-2jn\omega_0} - 3e^{-4nj\omega_0}]$$

$$= \frac{1}{6} [1 + 2(-1)^n - 3e^{-\frac{4}{3}jn\pi}]$$

$$X_n = \frac{1}{2in\pi} [1 + 2(-1)^n - 3e^{-\frac{4}{3}jn\pi}]$$

where

$$X_0 = \frac{1}{6} \int_{-1}^3 x(t) \ dt = 1$$

The plot for approximating the function using its Fourier coefficients is:

Listing 1.8: Approximation of x(t) with Fourier coefficients

```
function [x,t] = fapprox5(N,T)
t = -1.5*T:T/1000:1.5*T;
w0 = 2*pi/T;
X0 = 1;
```

```
5
       n_p = [1:N];
n_n = [-N:-1];
        Xn = (1 + 2*(-1).^n_p - 3*exp(-4*j*n_p*w0)) ./ (6*j*n_p*w0)
           );
        X_n = (1 + 2*(-1).^n_n - 3*exp(-4*j*n_n*w0)) ./ (6*j*n_n*
            w0);
10
       Xn = [X_n X0 Xn];
11
       n = [n_n \ 0 \ n_p];
12
13
        x = Xn*exp(j*w0*n'*t);
14
15
        x = real(x);
16 end
```

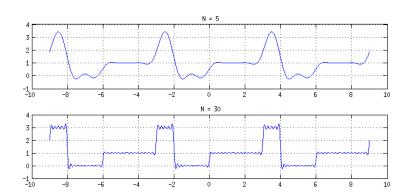


Figure 1.14: Approximation of x(t) by Xn

Chapter 2

Fourier Transform

When we want to know the description of aperiodic signals in terms of the frecuency content we need to use the Fourier Transform. The frecuency components of this non-periodic signals are defined for all reall values of the frecuency variable of ω and not just for discrete values as in the case of periodic ones in which we used the Fourier Series [1].

The Fourier Transform and its inverse of an aperiodic signal are defined as:

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

$$x(t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} X(\omega)e^{j\omega t} d\omega$$
(2.1)

In the next exercises we will also be using the following properties:

$$x(t) \Leftrightarrow X(\omega)$$
 (2.2a)

$$x(t-t_0) \Leftrightarrow e^{-j\omega t_0}X(\omega)$$
 (2.2b)

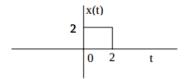
$$e^{j\omega_0 t}x(t) \Leftrightarrow X(\omega - \omega_0)$$
 (2.2c)

$$\frac{dx(t)}{dt} \Leftrightarrow j\omega X(\omega) \tag{2.2d}$$

$$\frac{dx(t)}{dt} \Leftrightarrow j\omega X(\omega)$$
 (2.2d)
$$x(\alpha t) \Leftrightarrow \frac{1}{|\alpha|} X\left(\frac{\omega}{a}\right)$$
 (2.2e)

For each signal, find the Fourier transform, $X(\omega)$, and then plot $|X(\omega)|$ (note, you may want to use MATLAB for the plot in 3.)

Problem 1



Solution

Taking the derivative of x(t) we get:

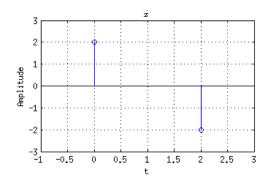


Figure 2.1: Derivative \dot{x}

Applying 2.1 and 2.2c to \dot{x} we have:

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

$$= 2[1 - e^{-2j\omega}]$$

$$= 2e^{-j\omega}[e^{-j\omega} - e^{-j\omega}]$$

$$= 4je^{-j\omega}\sin(\omega)$$

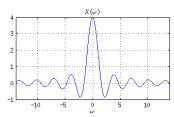
$$X(\omega) = 4\frac{\sin(\omega)}{\omega}e^{-j\omega}$$

$$= 4Sa(\omega)e^{-j\omega}$$

The plot of the magintude and angle of $X(\omega)$ is:

Listing 2.1: Plot of Magnitude and Angle

```
1 LIM = 3*pi;
2 w = -1.5*LIM:LIM/1000:1.5*LIM;
3 Xw = 4 * sin(w)./w .* exp(-j * w);
4
5 subplot(1,2,1), plot(w,real(Xw));
6 grid on;
7 xlim([-round(1.5*LIM) round(1.5*LIM)]);
8 xlabel('$\omega$','interpreter','latex');
10 title('$|X(\omega)|$','interpreter','latex');
10 subplot(1,2,2), plot(w,angle(Xw));
11 grid on;
12 xlim([-round(1.5*LIM) round(1.5*LIM)]);
13 xlabel('$\omega$','interpreter','latex');
14 title('$\ample X(\omega)$','interpreter','latex');
```



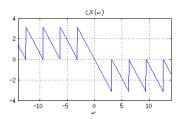
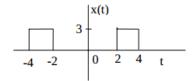


Figure 2.2: Magnitude $|X(\omega)|$ and Angle



Solution

Lets call $x_{p1}(t)$ and $X_{p1}(\omega)$ the function in time domain of the first point and its Fourier Transform respectively. We can express our function in terms of such function as:

$$x(t) = \frac{3}{2}[x_{p1}(t-2) + x_{p1}(t+4)]$$

As we can see from (2.2b) the displacement in time is reflected in the frecuency domain as a multiplication by an exponential.

$$X(\omega) = \frac{3}{2} X_{p1}(\omega) [e^{-2j\omega} + e^{4j\omega}]$$
$$= \frac{3}{2} X_{p1}(\omega) e^{j\omega} [e^{-3j\omega} + e^{3j\omega}]$$
$$= 12 Sa(\omega) \cos(3\omega)$$

The plot of the magintude and angle of $X(\omega)$ is:

Listing 2.2: Plot of Magnitude and Angle

```
1 LIM = 1.5*pi;
2 w = -1.5*LIM:LIM/1000:1.5*LIM;
3 Xw = 12 * sin(w)./w .* cos(3*w);
4
5 subplot(1,2,1), plot(w,abs(real(Xw)));
6 grid on;
7 xlim([-round(1.5*LIM) round(1.5*LIM)]);
8 xlabel('$\omega$','interpreter','latex');
10 title('$|X(\omega)|$','interpreter','latex');
10
11 subplot(1,2,2), plot(w,angle(Xw));
```

```
12 grid on;
13 xlim([-round(1.5*LIM) round(1.5*LIM)]);
14 xlabel('$\omega$','interpreter','latex');
15 title('$\angle X(\omega)$','interpreter','latex');
```

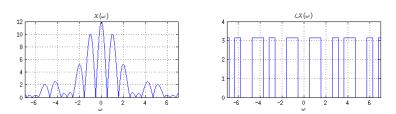
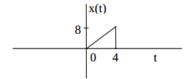


Figure 2.3: Magnitude $|X(\omega)|$ and Angle



Solution

Taking the derivative of x(t) we get:

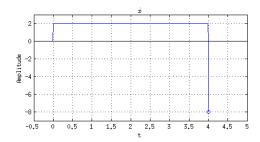


Figure 2.4: Derivative \dot{x}

Lets call $x_{p1}(t)$ and $X_{p1}(\omega)$ the function in time domain of the first point and its Fourier Transform respectively. We can express the first derivative of our function in terms of such function as:

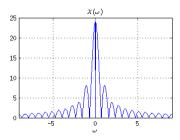
$$\dot{x}(t) = x_{p1}(t/2) - 8\delta(t-4)$$

As we can see from (2.2e) the scale in time is reflected inversely in the frecuency domain.

$$j\omega X(\omega) = X_{p1}(2\omega) - 8e^{-4j\omega}$$
$$= 4Sa(2\omega)e^{-2j\omega} - 8e^{-4j\omega}$$
$$X(\omega) = \frac{4j}{\omega}e^{-2j\omega}[2e^{-2j\omega} - Sa(2\omega)]$$

The plot of the magintude and angle of $X(\omega)$ is:

```
Listing 2.3: Plot of Magnitude and Angle
```



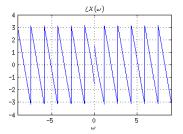


Figure 2.5: Magnitude $|X(\omega)|$ and Angle

$$x(t) = \cos(200t)G_4(t)$$

Solution

The generic Fourier Transform for the gate function $G_{\tau}(t)$ is:

$$\mathfrak{F}\{G_{\tau}(t)\} = \tau Sa\left(\frac{\omega\tau}{2}\right)$$

$$\mathfrak{F}\{G_{4}(t)\} = 4Sa(2\omega)$$
(2.3)

We can express our function as:

$$x(t) = \frac{1}{2} [e^{200t} + e^{-200t}] G_4(t)$$

As we can see from (2.2c) the displacement in frecuency is reflected in the time domain as a multiplication by an exponential.

$$X(\omega) = 2[Sa(2(\omega - 200)) + Sa(2(\omega + 200))]$$

The plot of the magintude of $X(\omega)$ is:

```
1  Sa=@(x) sin(x)./x;
2
3  LIM = 100*pi;
4  w = -1.5*LIM:LIM/1000:1.5*LIM;
5  Xw = 2 * ( Sa(2*(w-200)) + Sa(2*(w+200)) );
6
7  plot(w, abs(real(Xw)));
8  grid on;
9  xlim([-round(1.5*LIM) round(1.5*LIM)]);
10  xlabel('$\omega$','interpreter','latex');
```

title('\$|X(\omega)|\$','interpreter','latex');

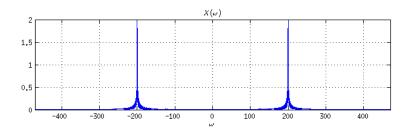


Figure 2.6: Magnitude $|X(\omega)|$

$$x(t) = e^{-3t}cos(10t)u(t)$$

Solution

From the 1st result of the Fourier Transforms table we have:

$$\mathfrak{F}\{e^{-\alpha t}u(t)\} = \frac{1}{\alpha + j\omega} \tag{2.4}$$

Rearranging the equation for x(t) we have:

$$x(t) = \frac{1}{2} [e^{-3t}e^{10jt} + e^{-3t}e^{-10jt}]u(t)$$
$$= \frac{1}{2} [e^{-t(3-10j)} + e^{-t(3+10j)}]u(t)$$

We can now apply directly the result from (2.5):

$$X(\omega) = \frac{1}{2} \left[\frac{1}{(3 - 10j) + j\omega} + \frac{1}{(3 + 10j) + j\omega} \right]$$
$$= \frac{1}{2} \left[\frac{1}{3 + (\omega - 10)j} + \frac{1}{3 + (\omega + 10)j} \right]$$

The plot of the magintude of $X(\omega)$ is:

Listing 2.5: Plot of Magnitude

```
1 LIM = 10*pi;
2 w = -1.5*LIM:LIM/1000:1.5*LIM;
3 Xw = 0.5 * ( 1./(3+(w-10)*j) + 1./(3+(w+10)*j));
4
5 plot(w,abs(real(Xw)));
6 grid on;
7 xlim([-round(1.5*LIM) round(1.5*LIM)]);
8 xlabel('$\omega$','interpreter','latex');
9 title('$\|X(\omega)\|$','interpreter','latex');
```

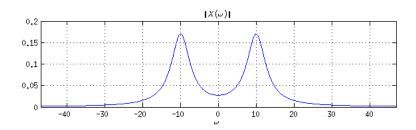
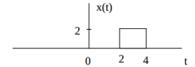


Figure 2.7: Magnitude $|X(\omega)|$

Problem 6

Find the Fourier transform of the following signals. Sketch $|X(\omega)|$ in each case.

1.:



Solution

Lets call $x_{p1}(t)$ and $X_{p1}(\omega)$ the function in time domain of the first point and its Fourier Transform respectively. We can express our function in terms of such function as:

$$x(t) = x_{p1}(t-2)$$

As we can see from (2.2b) the displacement in time is reflected in the frecuency domain as a multiplication by an exponential.

$$X(\omega) = X_{p1}(\omega)e^{-2j\omega}$$
$$= 4Sa(\omega)e^{-3j\omega}$$

The plot of the magintude of $X(\omega)$ is:

Listing 2.6: Plot of Magnitude

```
1  Sa=@(x) sin(x)./x;
2
3  LIM = -0.8*pi;
4  w = -1.5*LIM:LIM/1000:1.5*LIM;
5  Xw = 4 * Sa(w) .* exp(-3*j*w);
6
7  plot(w, abs(real(Xw)));
8  grid on;
9  xlim([-round(1.5*LIM) round(1.5*LIM)]);
10  xlabel('$\omega$','interpreter','latex');
11  title('$|X(\omega)|$','interpreter','latex');
```

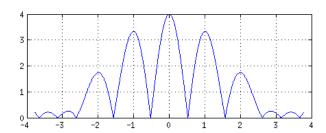


Figure 2.8: Magnitude $|X(\omega)|$

2.
$$x(t) = 2e^{-2t}u(t)$$

Solution

Directly from (2.5) we have:

$$X(\omega) = \frac{2}{2 + j\omega}$$

The plot of the magintude of $X(\omega)$ is:

Listing 2.7: Plot of Magnitude

```
1 LIM = 2*pi;
2 w = -1.5*LIM:LIM/1000:1.5*LIM;
3 Xw = 2./(2 + j*w);
4
5 plot(w,abs(real(Xw)));
6 grid on;
7 xlim([-round(1.5*LIM) round(1.5*LIM)]);
8 xlabel('$\omega$','interpreter','latex');
9 title('$|X(\omega)|$','interpreter','latex');
```

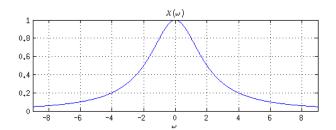


Figure 2.9: Magnitude $|X(\omega)|$

3.
$$x(t) = 5e^{-5t}u(t)$$

Solution

Directly from (2.5) we have:

$$X(\omega) = \frac{5}{5 + j\omega}$$

The plot of the magintude of $X(\omega)$ is:

Listing 2.8: Plot of Magnitude

```
1 LIM = 3*pi;
2 w = -1.5*LIM:LIM/1000:1.5*LIM;
3 Xw = 5./(5 + j*w);
4
5 plot(w,abs(real(Xw)));
6 grid on;
7 xlim([-round(1.5*LIM) round(1.5*LIM)]);
8 xlabel('$\omega$','interpreter','latex');
9 title('$|X(\omega)|$','interpreter','latex');
```

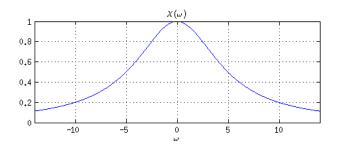


Figure 2.10: Magnitude $|X(\omega)|$

4.
$$x(t) = e^{-2t} \cos(4t)u(t)$$

Solution

$$x(t) = \frac{1}{2} \left[e^{-2t} e^{4jt} u(t) + e^{-2t} e^{-4jt} u(t) \right]$$

Directly from (2.5) we have:

$$X(\omega) = \frac{1}{2} \left[\frac{1}{(2-4j)+j\omega} + \frac{1}{(2+4j)+j\omega} \right]$$
$$= \frac{1}{2} \left[\frac{1}{2+(\omega-4)j} + \frac{1}{2+(\omega+4)j} \right]$$

The plot of the magintude of $X(\omega)$ is:

Listing 2.9: Plot of Magnitude

```
1 LIM = 2*pi;
2 w = -1.5*LIM:LIM/1000:1.5*LIM;
```

```
3  Xw = 0.5* (1./(2 + (w-4)*j) + 1./(2 + (w+4)*j));
4
5  plot(w,abs(real(Xw)));
6  grid on;
7  xlim([-round(1.5*LIM) round(1.5*LIM)]);
8  xlabel('$\omega$','interpreter','latex');
9  title('$|X(\omega)|$','interpreter','latex');
```

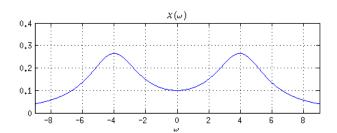
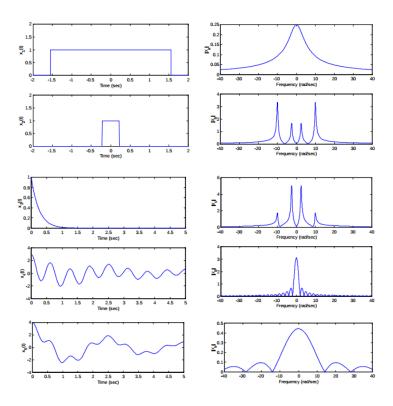


Figure 2.11: Magnitude $|X(\omega)|$

Match the time responses with the corresponding frequency responses.



Solution

As stated in (2.3) the Fourier Transform of a Gate function is a Sa. The 1st gate is more spread in time that the 2nd and therefore its corresponding transform must be more narrow.

 $1 \Leftrightarrow d$

 $2 \Leftrightarrow e$

The 3rd plot corresponds to a negative exponential truncated by a step function. We know for (2.5) and for Points 6b and 6c that the corresponding transform corresponds to a figure like (a).

 $3 \Leftrightarrow a$

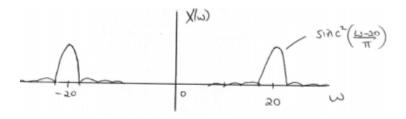
From the following plots we see that the frecuency of 5 (rad/s) is more dominant in the last one and the 10 (rad/s) is more dominant in the previous one.

 $4 \Leftrightarrow b$

 $5 \Leftrightarrow c$

Problem 8

Compute the inverse Fourier transform of the following signal



Solution

From the 14th result of the Fourier Transforms table we have:

$$Tr_{ au}(t) = \begin{cases} 1 - rac{|t|}{ au}, & \text{if } |t| < \tau \\ 0, & \text{if } |t| > \tau \end{cases}$$

$$\mathfrak{F}\{Tr_{\tau}(t)\} = \tau \left[Sa\left(\frac{\omega\tau}{2}\right) \right]^{2}$$

$$\mathfrak{F}\{Tr_{2}(t)\} = 2Sa(\omega)^{2}$$
(2.5)

First, we transform sinc into a known function like Sa:

$$sinc^2\left(\frac{\omega}{\pi}\right) = Sa^2(\omega)$$

As we saw in Point 4, when we multiply by a \cos in time we are adding 2 components of the function displaced simetrically with half of its amplitude. The displacement in the plot is of 20, therefore we need to multiply by a $\cos(20t)$. We end up having:

$$x(t) = Tr_2(t)\cos(20t)$$

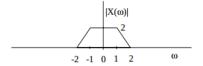
Chapter 3

Sampling and Reconstruction

Problem 1

Draw $|Xs(\omega)|$ for the following cases if $x_s(t)=x(t)p(t)$ with sampling period T.

$$p(t) = \sum_{n = -\infty}^{n = \infty} \delta(t - nT).$$



Solution

The minimum sampling period of the signal according to the Nyquist theorem is:

$$\omega_m = 2$$

$$\omega_{Ns} >= 2\omega_m >= 4$$

$$T_{Ns} = \frac{2\pi}{\omega_s} <= \frac{\pi}{2}$$

• $T_s = \pi/4sec$

This sampling period is lower than the minimum required and therefore no aliasing will occur as can be seen in (3.1).

$$\omega_s = \frac{2\pi}{T_s} = 8$$

Listing 3.1: Plot of $|X_s(\omega)|$

```
1 t1=[-2:0.01:-1];
2 x1=[2*t1+4];
3 t2=[-1:0.01:1];
4 x2=[0*t2+2];
5 t3=[1:0.01:2];
6 x3=[-2*t3+4];
7
8 t=[t1 t2 t3];
9 x=[x1 x2 x3];
10
11 ws = 8;
12 t = [t-ws t t+ws];
13 x = [x x x];
14
15 plot(t,x);
16 grid on;
17 ylim([-0.5 2.5]);
18 xlabel('$\omega$','interpreter','latex');
19 title('$|X_s(\omega)|$','interpreter','latex');
```

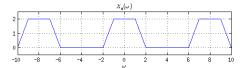


Figure 3.1: Sampling $|X_s(\omega)|$

• $T_s = \pi/2sec$

This sampling period is equal than the minimum required and therefore is in the limit of no aliasing as can be seen in (3.2).

$$\omega_s = \frac{2\pi}{T_s} = 4$$

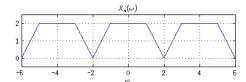


Figure 3.2: Sampling $|X_s(\omega)|$

•
$$T_s = 2\pi/3sec$$

This sampling period is lower than the minimum required and therefore aliasing will occur as can be seen in (3.3).

$$\omega_s = \frac{2\pi}{T_s} = 4$$

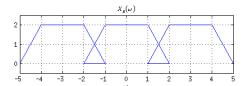


Figure 3.3: Sampling $|X_s(\omega)|$

Problem 2

Repeat Problem 1 where $x(t) = e^{-t/4}cos(t)u(t)$

In order to examine the effects of aliasing in the time domain, plot x(t) for each of the sampling times for t=0 to 15 sec. In MATLAB, this is done by defining your time vector with the time increment set to the desired sampling period. MATLAB then "reconstructs" the signal by connecting the sampled points with straight lines (this is known as a linear interpolation). Compare your sampled/reconstructed signals with a signal that is more accurate, one that is created by using a very small sampling period (such as T=0.05 sec) by plotting them on the same graph.

First we calculate $X(\omega)$.

$$x(t) = \frac{1}{2} \left[e^{-t/4} e^{jt} u(t) + e^{-t/4} e^{-jt} u(t) \right]$$

Directly from (2.5) we have:

$$X(\omega) = \frac{1}{2} \left[\frac{1}{(1/4+j)+j\omega} + \frac{1}{(1/4-j)+j\omega} \right]$$
$$= \frac{1}{2} \left[\frac{1}{1/4+(\omega+1)j} + \frac{1}{1/4+(\omega-1)j} \right]$$

The plot of the magintude of $X(\omega)$ is:

```
1 LIM = 0.5*pi;
2 w = -1.5*LIM:LIM/1000:1.5*LIM;
3 Xw = 0.5* (1./(0.25 + (w-1)*j) + 1./(0.25 + (w+1)*j));
4
5 plot(w,abs(real(Xw)));
6 grid on;
7 xlim([-round(1.5*LIM) round(1.5*LIM)]);
8 xlabel('$\omega$','interpreter','latex');
9 title('$|X(\omega)|$','interpreter','latex');
```

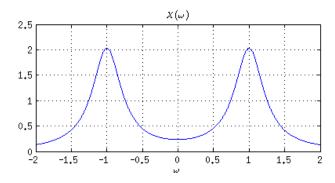


Figure 3.4: Magnitude $|X(\omega)|$

• $T_s = \pi/4sec$

This sampling period is lower than the minimum required and therefore no aliasing will occur as can be seen in (3.5).

$$\omega_s = \frac{2\pi}{T_s} = 8$$

Listing 3.3: Plot of $|X_s(\omega)|$

```
1 LIM = 2.5*pi;
2 w = -1.5*LIM:LIM/1000:1.5*LIM;
3 Xw = 0.5* (1./(0.25 + (w-1)*j) + 1./(0.25 + (w+1)*j));
4 Xw = abs(real(Xw));
5
6 ws = 8;
7 w = [w-ws w w+ws];
8 Xw = [Xw Xw Xw];
9
10 plot(w,Xw);
11 grid on;
12 xlim([-round(1.5*LIM) round(1.5*LIM)]);
13 ylim([-0.5 2.5]);
14 xlabel('$\omega$','interpreter','latex');
15 title('$|X_s(\omega)|$','interpreter','latex');
```

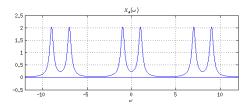


Figure 3.5: Sampling $|X_s(\omega)|$

•
$$T_s = \pi/2sec$$

This sampling period is equal than the minimum required and therefore is in the limit of no aliasing as can be seen in (3.6).

$$\omega_s = \frac{2\pi}{T_s} = 4$$

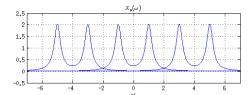


Figure 3.6: Sampling $|X_s(\omega)|$

•
$$T_s = 2\pi/3sec$$

This sampling period is lower than the minimum required and therefore aliasing will occur as can be seen in (3.7).

$$\omega_s = \frac{2\pi}{T_s} = 4$$

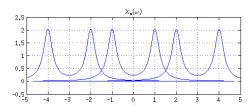


Figure 3.7: Sampling $|X_s(\omega)|$

Plot of "reconstructed" x(t) with MATLAB:

Listing 3.4: Plot of x(t) for different

```
1    u = @(x)(x>=0);
2
3    i=1;
4    for T=[0.05, pi/4, pi/2, 2*pi/3]
5     t=[0:T:15];
6     x=exp(-t/4).*cos(t).*u(t);
7
8     subplot(4,1,i), plot(t,x);
9     grid on;
10     xlabel('t');
11     title(['Reconstructed x(t) with T=' num2str(T)]);
12    i = i + 1;
13    end
```

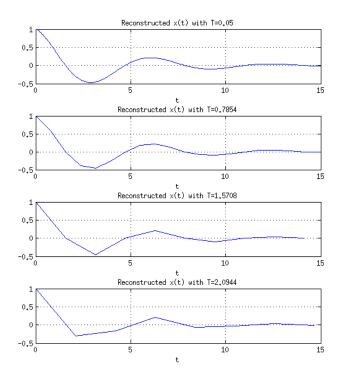
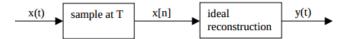


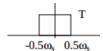
Figure 3.8: Reconstructed x(t)

Problem 3

Consider the following sampling and reconstruction configuration:



The output y(t) of the ideal reconstruction can be found by sending the sampled signal $x_s(t)=x(t)p(t)$ through an ideal lowpass filter:



Let $x(t) = 2 + cos(50\pi t)$ and T = 0.01 sec.

• Draw $|X_s(\omega)|$ where $x_s(t) = x(t)p(t)$. Determine if aliasing occurs.

Solution

First we calculate $X(\omega)$.

$$x(t) = 2(1) + \frac{1}{2} \left[e^{50\pi jt}(1) + e^{-50\pi jt}(1) \right]$$

Using this known result $\mathfrak{F}\{1\}=2\pi\delta(\omega)$ and from (2.2c) we have:

$$X(\omega) = 4\pi\delta(\omega) + \pi \left[\delta(\omega - 50\pi) + \delta(\omega + 50\pi)\right]$$

The plot of the magintude of $X(\omega)$ is:

```
Listing 3.5: Plot of Magnitude
```

```
1  fe = [-50*pi 0 50*pi];
2  Ae = [pi 4*pi pi];
3
4  figure(1);
5  stem(fe,Ae);
6  grid on;
7  xlabel('Frec [rad/s]');
8  ylabel('Amplitude');
9  title('|X_\omega|', 'fontweight', 'bold');
```

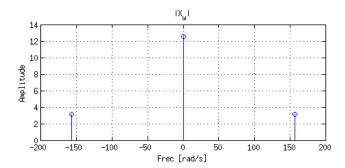


Figure 3.9: Magnitude $|X(\omega)|$

The minimum sampling period of the signal according to the Nyquist theorem is:

$$\begin{aligned} \omega_m &= 50\pi\\ \omega_{Ns}>&= 2\omega_m> = 100\pi\\ T_{Ns} &= \frac{2\pi}{\omega_s} <= \frac{1}{50} <= 0.02 \end{aligned}$$

Since $T_s = 0.01sec$ is lower than the minimum required no aliasing will occur as can be seen in (3.10).

$$\omega_s = \frac{2\pi}{T_s} = 200\pi$$

```
Listing 3.6: Plot of |X_s(\omega)|
```

```
1  w = [-50*pi 0 50*pi];
2  Xw = [pi 4*pi pi];
3
4  ws = 200*pi;
5  w = [w-ws w w+ws];
6  Xw = [Xw Xw Xw];
7
8  stem(w,Xw);
9  grid on;
10  %*xlim([-50*pi-2 50*pi+2]);
11  xlabel('$\omega$','interpreter','latex');
12  title('$|X_s(\omega)|$','interpreter','latex');
```

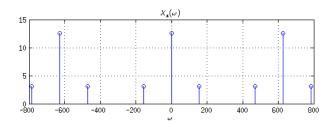


Figure 3.10: Sampling $|X_s(\omega)|$

• Determine the expression for y(t).

Solution

The limits of the lowpass filter are $-0.5\omega_s=-100\pi$ to $0.5\omega_s=100\pi$. The Fourier transform of $Y(\omega)=H(\omega)X(\omega)$ is:

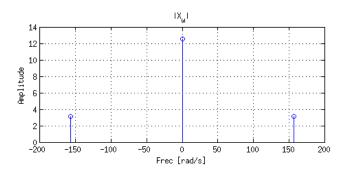


Figure 3.11: Magnitude $|X(\omega)|$

And therefore:

$$y(t) = 2 + \cos(50\pi t)$$

• Determine an expression for x[n].

$$x(n) = 2 + \cos(50\pi nT_s)$$
$$= 2 + \cos(0.5\pi n)$$

Problem 4

Repeat Problem 3 for $x(t) = 2 + \cos(50\pi t)$ and T = 0.025 sec.

• Draw $|X_s(\omega)|$ where $x_s(t) = x(t)p(t)$. Determine if aliasing occurs.

Solution

Since $T_s=0.025sec$ is greater than the minimum required, aliasing will occur as can be seen in (3.12).

$$\omega_s = \frac{2\pi}{T_s} = 80\pi$$

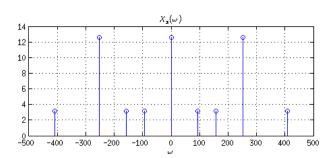


Figure 3.12: Sampling $|X_s(\omega)|$

• Determine the expression for y(t).

The limits of the lowpass filter are $-0.5\omega_s=-40\pi$ to $0.5\omega_s=40\pi$. The Fourier transform of $Y(\omega)=H(\omega)X(\omega)$ is:

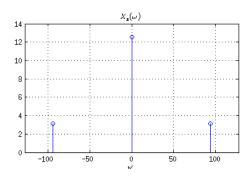


Figure 3.13: Magnitude $|X(\omega)|$

And therefore:

$$y(t) = 2 + \cos(30\pi t)$$

• Determine an expression for x[n].

Solution

$$x(n) = 2 + \cos(50\pi nT_s)$$
$$= 2 + \cos(1.25\pi n)$$

Problem 5

Repeat Problem 3 for $x(t) = 1 + \cos(20\pi t) + \cos(60\pi t)$ and T = 0.01 sec.

• Draw $|X_s(\omega)|$ where $x_s(t)=x(t)p(t)$. Determine if aliasing occurs.

First we calculate $X(\omega)$.

$$X(\omega) = \pi[2\delta(\omega) + \delta(\omega - 20\pi) + \delta(\omega + 20\pi) + \delta(\omega - 60\pi) + \delta(\omega + 60\pi)]$$

The plot of the magintude of $X(\omega)$ is:

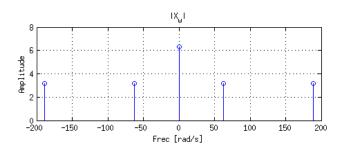


Figure 3.14: Magnitude $|X(\omega)|$

The minimum sampling period of the signal according to the Nyquist theorem is:

$$\omega_m = 60\pi$$

$$\omega_{Ns} >= 2\omega_m >= 120\pi$$

$$T_{Ns} = \frac{2\pi}{\omega_s} <= \frac{1}{50} <= 0.01666$$

Since $T_s = 0.01sec$ is lower than the minimum required no aliasing will occur as can be seen in (3.15).

$$\omega_s = \frac{2\pi}{T_s} = 200\pi$$

Listing 3.8: Plot of $|X_s(\omega)|$

```
1  w = [-60*pi -20*pi 0 20*pi 60*pi];
2  Xw = [pi pi 2*pi pi pi];
3
4  ws = 200*pi;
5  w = [w-ws w w+ws];
6  Xw = [Xw Xw Xw];
7
8  figure(1);
9  stem(w,Xw);
10  grid on;
11  xlabel('Frec [rad/s]');
12  ylabel('Amplitude');
13  title('|X_\omega|', 'fontweight', 'bold');
```

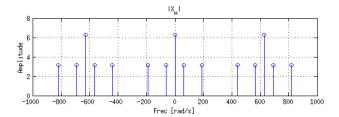


Figure 3.15: Sampling $|X_s(\omega)|$

• Determine the expression for y(t).

Solution

The limits of the lowpass filter are $-0.5\omega_s=-100\pi$ to $0.5\omega_s=100\pi$. The Fourier transform of $Y(\omega)=H(\omega)X(\omega)$ is:

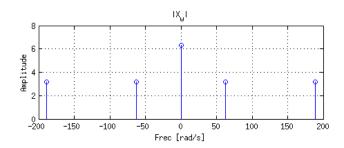


Figure 3.16: Magnitude $|X(\omega)|$

And therefore:

$$y(t) = x(t) = 1 + \cos(20\pi t) + \cos(60\pi t)$$

• Determine an expression for x[n].

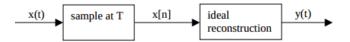
Solution

$$x(n) = 1 + \cos(20\pi nT_s) + \cos(60\pi nT_s)$$

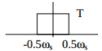
= 2 + \cos(0.2\pi n) + \cos(0.6n)

Problem 6

Consider the following sampling and reconstruction configuration:



The output y(t) of the ideal reconstruction can be found by sending the sampled signal $x_s(t)=x(t)p(t)$ through an ideal lowpass filter:



• Let $x(t)=1+cos(15\pi t)$ and T = 0.1 sec. Draw $|X_s(\omega)|$ where $x_s(t)=x(t)p(t)$. Determine the expression for y(t).

Solution

First we calculate $X(\omega)$.

$$X(\omega) = \pi[2\delta(\omega) + \delta(\omega - 15\pi) + \delta(\omega + 15\pi)]$$

The plot of the magintude of $X(\omega)$ is:

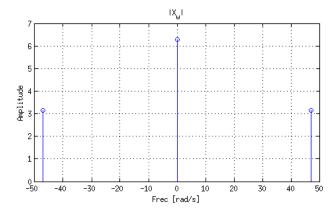


Figure 3.17: Magnitude $|X(\omega)|$

The minimum sampling period of the signal according to the Nyquist theorem is:

$$\begin{aligned} \omega_m &= 15\pi\\ \omega_{Ns} >&= 2\omega_m > = 30\pi\\ T_{Ns} &= \frac{2\pi}{\omega_s} <= \frac{1}{15} <= 0.066 \end{aligned}$$

Since $T_s=0.1sec$ is greater than the minimum required, aliasing will occur as can be seen in (3.18).

$$\omega_s = \frac{2\pi}{T_s} = 20\pi$$

Listing 3.10: Plot of $|X_s(\omega)|$

```
1  w = [-15*pi 0 15*pi];
2  Xw = [pi 2*pi pi];
3
4  ws = 20*pi;
5  w = [w-ws w w+ws];
6  Xw = [Xw Xw Xw];
7
8  stem(w,Xw);
9  grid on;
10  %xlim([-10*pi-2 10*pi+2]);  %passband filter
11  xlabel('Free [rad/s]');
12  ylabel('Amplitude');
13  title('|x_\omega|', 'fontweight', 'bold');
```

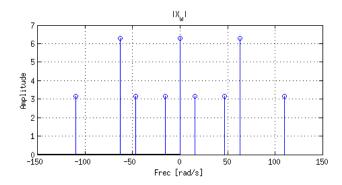


Figure 3.18: Sampling $|X_s(\omega)|$

The limits of the lowpass filter are $-0.5\omega_s=-10\pi$ to $0.5\omega_s=10\pi$. The

Fourier transform of $Y(\omega) = H(\omega)X(\omega)$ is:

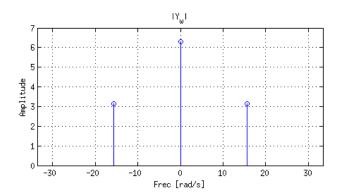


Figure 3.19: Magnitude $|X(\omega)|$

And therefore only the $\omega=5\pi$ passes through it:

$$y(t) = 1 + \cos(5\pi t)$$

• Let $X(\omega)=\frac{1}{(j\omega+1)}$ and T = 1 sec. Draw $|X_s(\omega)|$ where $x_s(t)=x(t)p(t)$. Does aliasing occur? (Justify your answer).

Solution

The magnitude of $X(\omega)$ is:

$$|X(\omega)| = \frac{1}{\omega^2 + 1}$$

From the previous equation we can see that the function never intesects the w axis, and therefore there will always be aliasing. However, setting w_m to its FWHM we have:

$$\frac{1}{\omega_m^2+1}=\frac{1}{2}$$

$$\omega_m=\sqrt{3}$$

Hipoteticaly taking this value as w_m we see that $w_s=2\pi>2w_m=2\sqrt{3}$ and in this case there will not be aliasing.

The plot of the magintude of $X(\omega)$ and $X_s(\omega)$ is:

```
1 LIM = 1.5 * pi;
w = -1.5*LIM:LIM/1000:1.5*LIM;
  XW = 1./(j*w + 1);
  subplot (2,1,1), plot (w, abs (real (Xw)));
6 grid on;
   xlabel('$\omega$','interpreter','latex');
   title('$|X(\omega)|$','interpreter','latex');
10
11 hold on
12 stem(sqrt(3),1./(j*sqrt(3) + 1))
13
14 ws = 2 * pi;
15 W = [W-WS W W+WS];
16 XW = [XW XW XW];
17
   subplot (2,1,2), plot (w, abs (real (Xw)));
   grid on;
19
   %xlim([-round(1.5*LIM) round(1.5*LIM)]);
   xlabel('$\omega$','interpreter','latex');
21
   title('$|X_s(\omega)|$','interpreter','latex');
22
24 %FWHM
25 hold on
26 stem(sqrt(3),1./(j*sqrt(3) + 1))
```

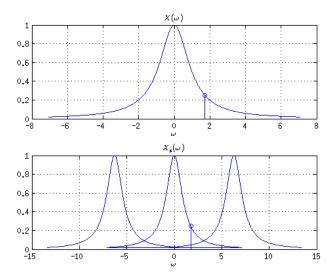


Figure 3.20: Magnitude

Chapter 4

DTFT and **DFT**

When we move into a discrete time domain, the Fourier Transform equivalent in this domain is called the DTFT or Discrete Time Fourier Transform which is defined as [1]:

$$\mathfrak{F}\{x[n]\} = \sum_{-\infty}^{\infty} x[n]e^{-j\omega n} = X(\omega)$$
(4.1)

Problem 1

Compute the DTFT of the following signals and sketch $X(\omega)$.

•
$$x[n] = \begin{bmatrix} \frac{1}{4} & \frac{1}{4} & \frac{1}{4} & \frac{1}{4} \end{bmatrix}$$

Solution

Aplying the definition of (4.1) we have:

$$X(\omega) = \sum_{0}^{3} \frac{1}{4} e^{-j\omega n}$$

$$= \frac{1}{4} [1 + e^{-jw} + e^{-2jw} + e^{-3jw}]$$

$$= \frac{1}{4} e^{-j\frac{3\omega}{2}} [e^{j\frac{3\omega}{2}} + e^{-j\frac{\omega}{2}} + e^{-j\frac{\omega}{2}} + e^{-j\frac{3\omega}{2}}]$$

$$= \frac{1}{2} e^{-j\frac{3\omega}{2}} [\cos(\frac{3\omega}{2})\cos(\frac{\omega}{2})]$$

The plot of the magintude $X(\omega)$ is:

```
Listing 4.1: Plot of Magnitude
```

```
1 LIM = 2.3;
2 w = -1.5*LIM:LIM/1000:1.5*LIM;
3 Xw = 0.5 * exp(-j * 1.5 * w) .* ( cos(w*1.5) + cos(w*0.5) );
4
5 subplot(2,1,1), plot(w,abs(Xw));
6 grid on;
7 xlim([-round(1.5*LIM) round(1.5*LIM)]);
8 xlabel('$\omega$','interpreter','latex');
9 title('$\|X(\omega)\|$','interpreter','latex');
10
11 subplot(2,1,2), plot(w,angle(Xw));
12 grid on;
13 xlim([-round(1.5*LIM) round(1.5*LIM)]);
14 xlabel('$\omega$','interpreter','latex');
15 title('$\angle X(\omega)$','interpreter','latex');
```

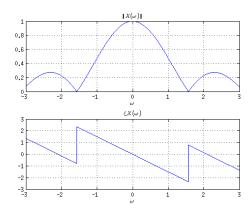


Figure 4.1: Magnitude $|X(\omega)|$

•
$$x[n] = \begin{bmatrix} 1 & -2 & 1 \end{bmatrix}$$

Aplying the definition of (4.1) we have:

$$X(\omega) = \sum_{0}^{2} x[n]e^{-j\omega n}$$

$$= [1 - 2e^{-jw} + e^{-2jw}]$$

$$= e^{-jw}[e^{jw} - 2 + e^{-jw}]$$

$$= e^{-jw})2[\cos(\omega) - 1]$$

$$= 2e^{-jw}[\cos(\omega) - 1]$$

The plot of the magintude $X(\omega)$ is:

```
Listing 4.2: Plot of Magnitude

1  LIM = 0.7*pi;
2  w = -1.5*LIM:LIM/1000:1.5*LIM;
3  Xw = 2 * exp(-j*w) .* (cos(w)-1);

4  subplot(2,1,1), plot(w,abs(Xw));
6  grid on;
7  xlim([-round(1.5*LIM) round(1.5*LIM)]);
8  xlabel('$\omega$','interpreter','latex');
9  title('$\|X(\omega)\|$','interpreter','latex');

10  subplot(2,1,2), plot(w,angle(Xw));
11  grid on;
12  xlim([-round(1.5*LIM) round(1.5*LIM)]);
13  xlabel('$\omega$','interpreter','latex');
14  title('$\ample X(\omega)$','interpreter','latex');
```

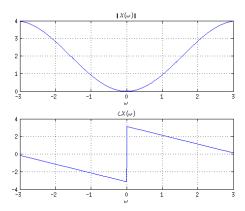


Figure 4.2: Magnitude $|X(\omega)|$

•
$$x[n] = 2\left(\frac{3}{4}\right)^n u[n]$$

$$X(\omega) = \sum_{0}^{\infty} 2\left(\frac{3}{4}\right)^{n} e^{-j\omega n}$$

Recall that the geometric series solution is given by:

$$\sum ar^n = \frac{a}{1-r} \text{ if } |r| < 1$$
 (4.2)

Then, aplying (4.1) and (4.2) we have:

$$X(\omega) = \sum_{0}^{\infty} 2\left(\frac{3}{4}e^{-j\omega}\right)^{n}$$
$$= \frac{2}{1 - \frac{3}{4}e^{-j\omega}}$$

The plot of the magintude $X(\omega)$ is:

Listing 4.3: Plot of Magnitude

1 LIM = 0.7 * pi;

```
2  w = -1.5*LIM:LIM/1000:1.5*LIM;
3  Xw = 2 ./(1 - 3/4*exp(-j*w));
4
5  subplot(2,1,1), plot(w,abs(Xw));
6  grid on;
7  xlim([-round(1.5*LIM) round(1.5*LIM)]);
8  xlabel('$\omega$','interpreter','latex');
10  subplot(2,1,2), plot(w,angle(Xw));
11  subplot(2,1,2), plot(w,angle(Xw));
12  grid on;
13  xlim([-round(1.5*LIM) round(1.5*LIM)]);
14  xlabel('$\omega$','interpreter','latex');
15  title('$\angle X(\omega)$','interpreter','latex');
```

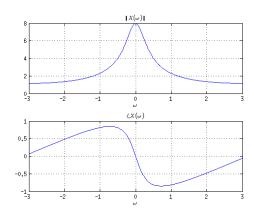


Figure 4.3: Magnitude $|X(\omega)|$

Chapter 5

Z-Transform

Chapter 6

Haar Base

Problem 1

Create a GUI that plots the approximation of the function $x(t)=t^2$ using Haar Base. The values for

$$\phi(2^j-t) = \begin{cases} 1, & \text{if } \frac{k}{2^j} <= t <= \frac{k+1}{2^j} \\ 0, & \text{in any other case} \end{cases}$$

With

$$C_k = 2^j \int_{\frac{k}{2^j}}^{\frac{k+1}{2^j}} f(t)\phi(t-k) dt$$

For j=0 we have:

$$\begin{split} C_k &= \int_k^{k+1} t^2 \phi(t-k) \ dt \\ &= \frac{1}{3} t^3 \bigg|_k^{k+1} \\ &= \frac{1}{3} [3k^2 + 3k + 1] \ \text{with} \ \ k = -3, -2, -1, 0, 1, 2 \\ f(t) &= \sum_{k=-3}^2 \frac{1}{3} [3k^2 + 3k + 1] \phi(t-k) \end{split}$$

For j¿=0 we use the following code to plot the GUI in matlab:

Listing 6.1: Approximation Function

```
function [ tk, fs ] = approximate_t2( j, t )
% APPROXIMATE Aproximacin de la funcin t^2 mediante steps

j2=2^j;
k=[j2*(-t) : j2*(t)-1]
fs = 1/(j2^2) * (k.^2 + k + 1/3)
tk = k/j2
% plot(tk,tk.^2,'r')
% hold on
% stairs(tk,fs,'b')

end
```

Listing 6.2: Plot of the Approximation Function

```
function [ output_args ] = plotAprox(tk,f,fs)
% PLOTAPROX Summary of this function goes here
% Detailed explanation goes here

x = linspace(tk(1),-tk(1),int32(20));
plot(x,f(x),'r')
hold on
stairs(tk,fs,'b')
end
```

Listing 6.3: GU

```
function varargout = haar(varargin)
HAAR MATLAB code for haar.fig
```

```
용
          HAAR, by itself, creates a new HAAR or raises the
3
        existing
   용
          singleton*.
4
   응
5
   용
          {\it H} = HAAR returns the handle to a new HAAR or the handle
6
        to
7
   응
          the existing singleton*.
   용
8
   용
          HAAR('CALLBACK', hObject, eventData, handles,...) calls the
         local
           function named CALLBACK in HAAR.M with the given input
10
        arguments.
11
12
  용
          HAAR ('Property', 'Value', ...) creates a new HAAR or
        raises the
   응
          existing singleton*. Starting from the left, property
13
        value pairs are
          applied to the GUI before haar_OpeningFcn gets called.
14
        An
   용
          unrecognized property name or invalid value makes
15
        property application
          stop. All inputs are passed to haar_OpeningFcn via
16
        varargin.
17
   용
   용
           *See GUI Options on GUIDE's Tools menu. Choose "GUI
18
        allows only one
   용
          instance to run (singleton)".
19
20
   % See also: GUIDE, GUIDATA, GUIHANDLES
21
22
23
   % Edit the above text to modify the response to help haar
24
   % Last Modified by GUIDE v2.5 04-Apr-2013 15:42:43
25
26
   % Begin initialization code - DO NOT EDIT
27
28
   gui_Singleton = 1;
   gui_State = struct('gui_Name',
                                          mfilename, ...
29
30
                       'gui_Singleton', gui_Singleton, ...
                       'gui_OpeningFcn', @haar_OpeningFcn, ...
31
32
                       'gui_OutputFcn', @haar_OutputFcn, ...
                       'gui_LayoutFcn',
33
                                          [],...
                       'gui_Callback',
34
                                          []);
   if nargin && ischar(varargin{1})
35
       gui_State.gui_Callback = str2func(varargin{1});
36
37
   end
38
39
        [varargout{1:nargout}] = gui_mainfcn(gui_State, varargin)
40
            {:});
41
   else
       gui_mainfcn(gui_State, varargin{:});
42
43
44
   % End initialization code - DO NOT EDIT
45
46
  function calcAndPlot(handles)
47
```

```
49 tmax = str2num(get(handles.tmax,'String'));
   j = str2num(get(handles.j,'String'));
   [tk,fs]=approximate_t2(j,tmax);
51
52 plotAprox(tk, handles.ft2, fs);
53
   % --- Executes just before haar is made visible.
54
55
  function haar_OpeningFcn(hObject, eventdata, handles, varargin)
   % This function has no output args, see OutputFcn.
  % hObject
               handle to figure
58
   % eventdata reserved - to be defined in a future version of
       MATLAB
59
   % handles
                structure with handles and user data (see GUIDATA)
   % varargin command line arguments to haar (see VARARGIN)
62 handles.ft2 = @(t) t.^2;
64 calcAndPlot(handles)
65
66
   % Set the current data value.
67
   %handles.current_data = handles.peaks;
   %plot (handles.current_data)
69
71 % Choose default command line output for haar
72 handles.output = hObject;
73
74 % Update handles structure
75 guidata(hObject, handles);
76
   % UIWAIT makes haar wait for user response (see UIRESUME)
77
78
   % uiwait (handles.figure1);
   % --- Outputs from this function are returned to the command
       line.
   function varargout = haar_OutputFcn(hObject, eventdata, handles
   % varargout cell array for returning output args (see
       VARARGOUT);
   % hObject
                handle to figure
   % eventdata reserved - to be defined in a future version of
       MATLAB
86
   % handles
                structure with handles and user data (see GUIDATA)
87
   % Get default command line output from handles structure
   varargout{1} = handles.output;
91
   % --- Executes on button press in pbCalcular.
92
   function pbCalcular_Callback(hObject, eventdata, handles)
               handle to pbCalcular (see GCBO)
94
   % hObject
   % eventdata reserved - to be defined in a future version of
       MATLAB
   % handles
                structure with handles and user data (see GUIDATA)
96
   calcAndPlot(handles)
  % --- Executes on selection change in popupmenu2.
```

```
function popupmenu2_Callback(hObject, eventdata, handles)
    % hObject
                 handle to popupmenu2 (see GCBO)
    % eventdata reserved - to be defined in a future version of
102
        MATLAB
    % handles
                 structure with handles and user data (see GUIDATA)
103
104
105
    % Hints: contents = cellstr(get(hObject,'String')) returns
        popupmenu2 contents as cell array
            contents{get(hObject,'Value')} returns selected item
106
        from popupmenu2
107
108
    % --- Executes during object creation, after setting all
109
        properties.
   function popupmenu2_CreateFcn(hObject, eventdata, handles)
110
    % hObject handle to popupmenu2 (see GCBO)
111
112
    % eventdata reserved - to be defined in a future version of
        MATLAB
    % handles
                 empty - handles not created until after all
113
        CreateFcns called
    % Hint: popupmenu controls usually have a white background on
115
        Windows.
116
           See ISPC and COMPUTER.
    if ispc && isequal(get(hObject,'BackgroundColor'), get(0,'
117
        defaultUicontrolBackgroundColor'))
        set(hObject,'BackgroundColor','white');
118
   end
119
120
121
    % --- Executes on slider movement.
122
123 function slider1_Callback(hObject, eventdata, handles)
                handle to slider1 (see GCBO)
    % hObject
   % eventdata reserved - to be defined in a future version of
125
        MATLAB
126
    % handles
                 structure with handles and user data (see GUIDATA)
127
   % Hints: get(hObject,'Value') returns position of slider
             get(hObject,'Min') and get(hObject,'Max') to determine
129
         range of slider
130
131
    % --- Executes during object creation, after setting all
        properties.
    function slider1_CreateFcn(hObject, eventdata, handles)
                handle to slider1 (see GCBO)
134
    % hObject
    % eventdata reserved - to be defined in a future version of
135
        {\it MATLAB}
    % handles
                 empty - handles not created until after all
136
        CreateFcns called
137
    % Hint: slider controls usually have a light gray background.
138
    if isequal(get(hObject,'BackgroundColor'), get(0,'
        defaultUicontrolBackgroundColor'))
140
        set(hObject,'BackgroundColor',[.9 .9 .9]);
   end
141
142
```

```
143
145 function tmax_Callback(hObject, eventdata, handles)
                handle to tmax (see GCBO)
146
    % hObject
   % eventdata reserved - to be defined in a future version of
147
        MATLAB
148
    % handles
                 structure with handles and user data (see GUIDATA)
149
   % Hints: get(hObject,'String') returns contents of tmax as text
150
            str2double(get(h0bject,'String')) returns contents of
151
        tmax as a double
152
153
    % --- Executes during object creation, after setting all
        properties.
    function tmax_CreateFcn(hObject, eventdata, handles)
155
156
    % hObject handle to tmax (see GCBO)
    % eventdata reserved - to be defined in a future version of
157
        {\it MATLAB}
    % handles
                 empty - handles not created until after all
158
        CreateFcns called
159
    % Hint: edit controls usually have a white background on
160
        Windows.
           See ISPC and COMPUTER.
161
    if ispc && isequal(get(hObject,'BackgroundColor'), get(0,'
162
        defaultUicontrolBackgroundColor'))
        set(hObject,'BackgroundColor','white');
163
   end
164
165
166
167
   function j_Callback(hObject, eventdata, handles)
168
    % hObject handle to j (see GCBO)
169
    % eventdata reserved - to be defined in a future version of
170
        MATLAB
    % handles
                 structure with handles and user data (see GUIDATA)
171
   % Hints: get(hObject,'String') returns contents of j as text
173
174
             str2double(get(hObject,'String')) returns contents of
        j as a double
175
176
   % --- Executes during object creation, after setting all
177
        properties.
178
   function j_CreateFcn(hObject, eventdata, handles)
    % hObject handle to j (see GCBO)
179
   % eventdata reserved - to be defined in a future version of
        MATTAB
    % handles
                 empty - handles not created until after all
        CreateFcns called
182
183
   % Hint: edit controls usually have a white background on
        Windows.
184
           See ISPC and COMPUTER.
   if ispc && isequal(get(hObject,'BackgroundColor'), get(0,'
185
        defaultUicontrolBackgroundColor'))
```

Chapter 7

Haar Transform

Bibliography

- [1] Edward Kamen and Bonnie Heck. *Fundamentals of Signals and Systems: With MATLAB Examples.* Prentice Hall PTR, 2000.
- [2] Georgia Tech School of Electrical and Computer Engineering. Worked Problems, Chapter 4. http://users.ece.gatech.edu/~bonnie/book/worked_problems/Chap4_Fseries_sol.pdf. [Online; accessed Apr-2013].