

Lab-5

Discrete time FT and LTI systems

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1.Discrete time Fourier transform(DTFT):

For finding the DTFT of any discrete time signal $x[n]$ is given as

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

Here the function

$X = \text{DT_Fourier}(x, N0, w)$

CODE:

```
1 function X = DT_Fourier(x, N0, w)
2     N = length(x);
3     X = zeros(size(w));
4     for i = 1:length(w)
5         X(i) = sum(x .* exp(-1j * w(i) * (0:N-1 - N0)));
6     end
7 end
```

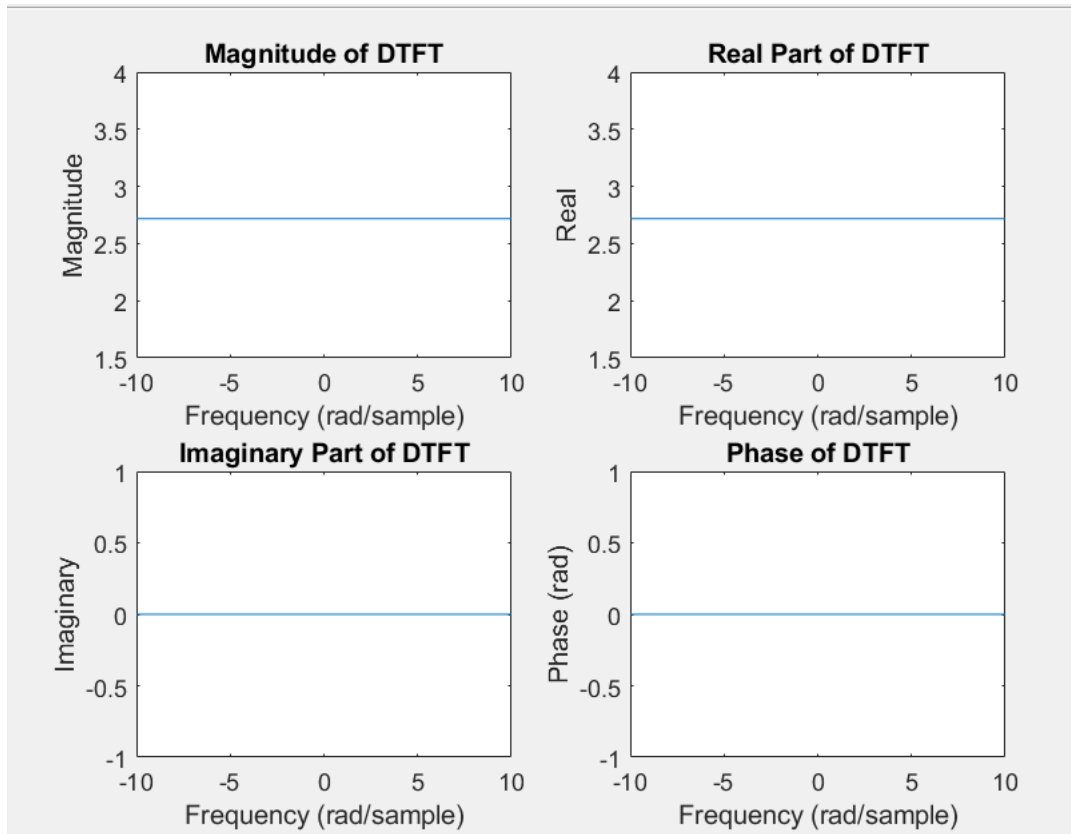
Here

- x , a discrete-time signal of finite duration (assume that the signal is zero elsewhere)
- $N0$, location of the sample $x[0]$ in the given input signal x , note that $N0$ is a positive integer between 1 and $\text{length}(x)$
- ω , a vector of frequencies at which to compute the DTFT (though frequency is a continuous variable in DTFT we can evaluate it at only finite set of points)

b) Set $\omega = -10: 0.01: 10$

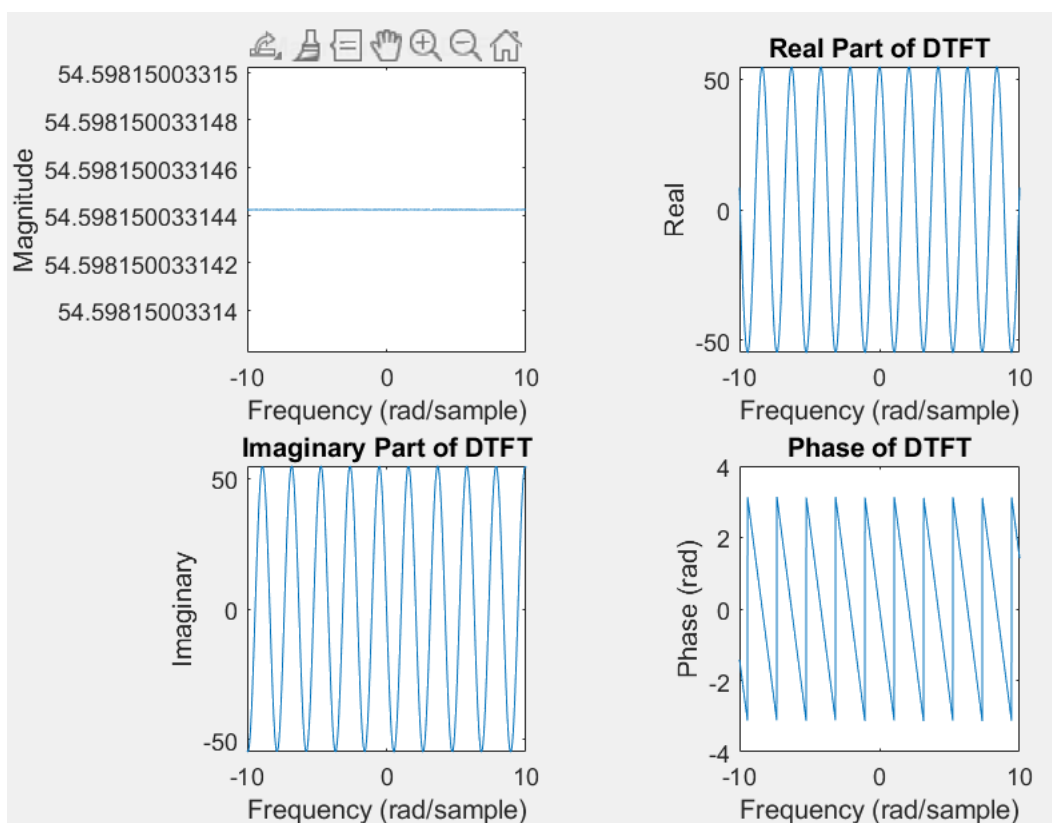
1) unit impulse $\delta[n]$

Plot:



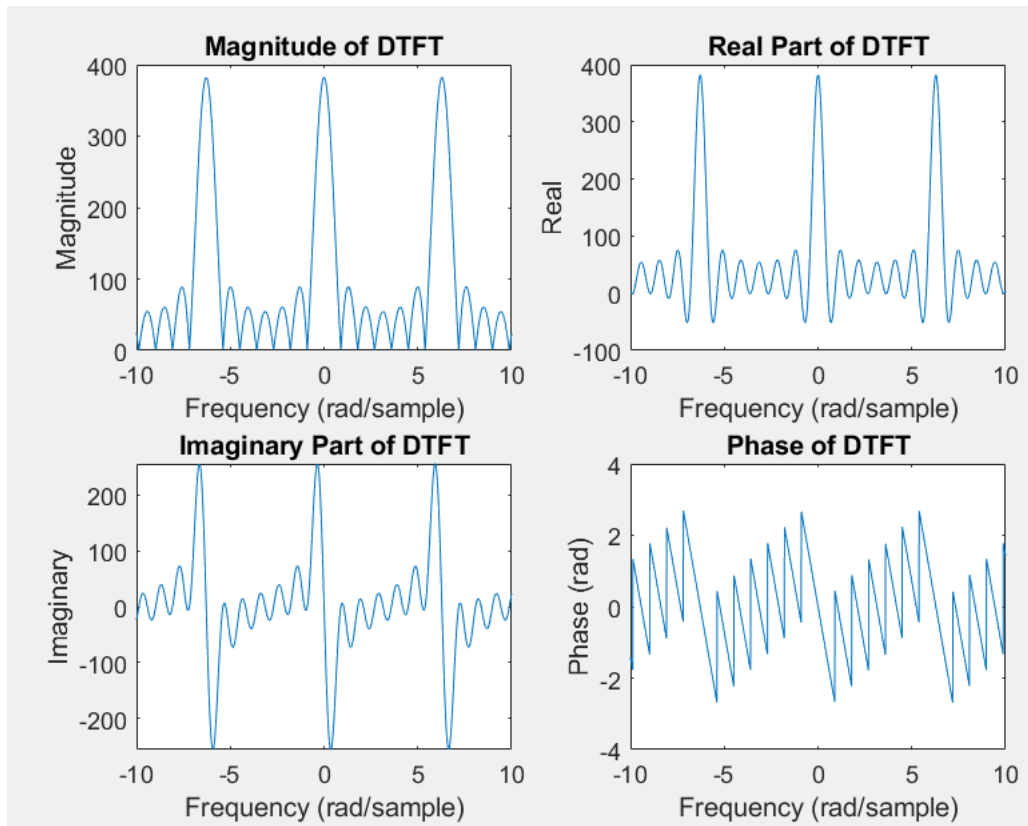
b2)

Plot:



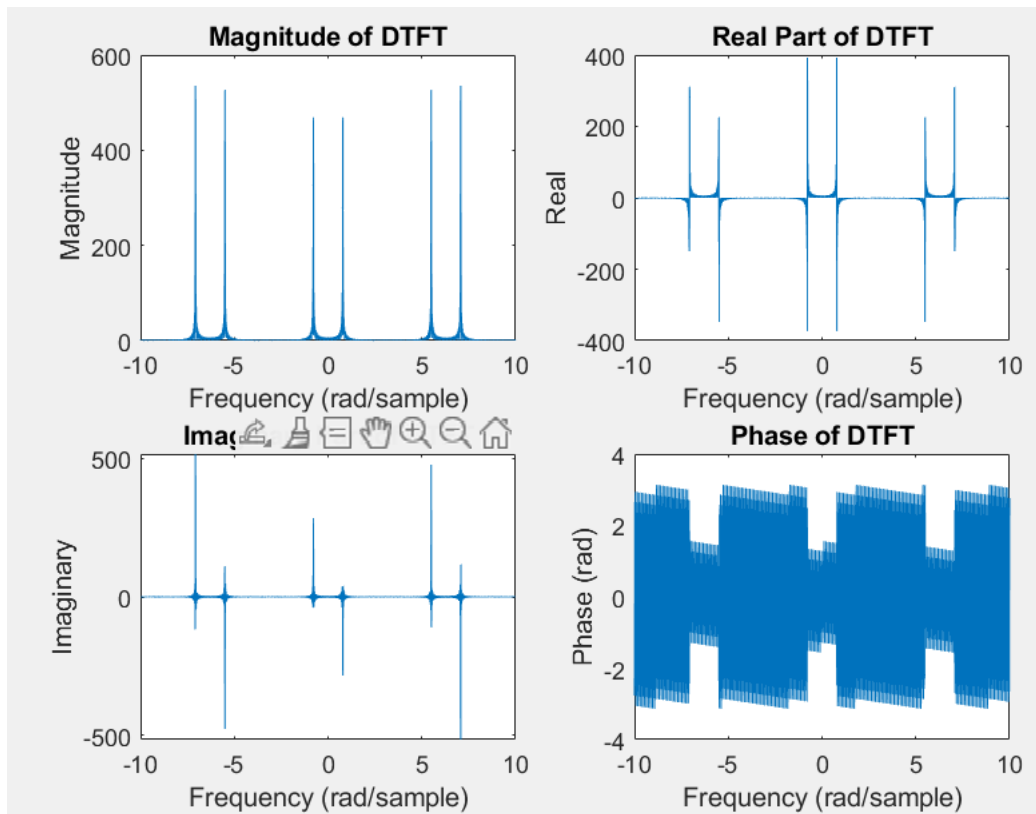
b3)

plot:

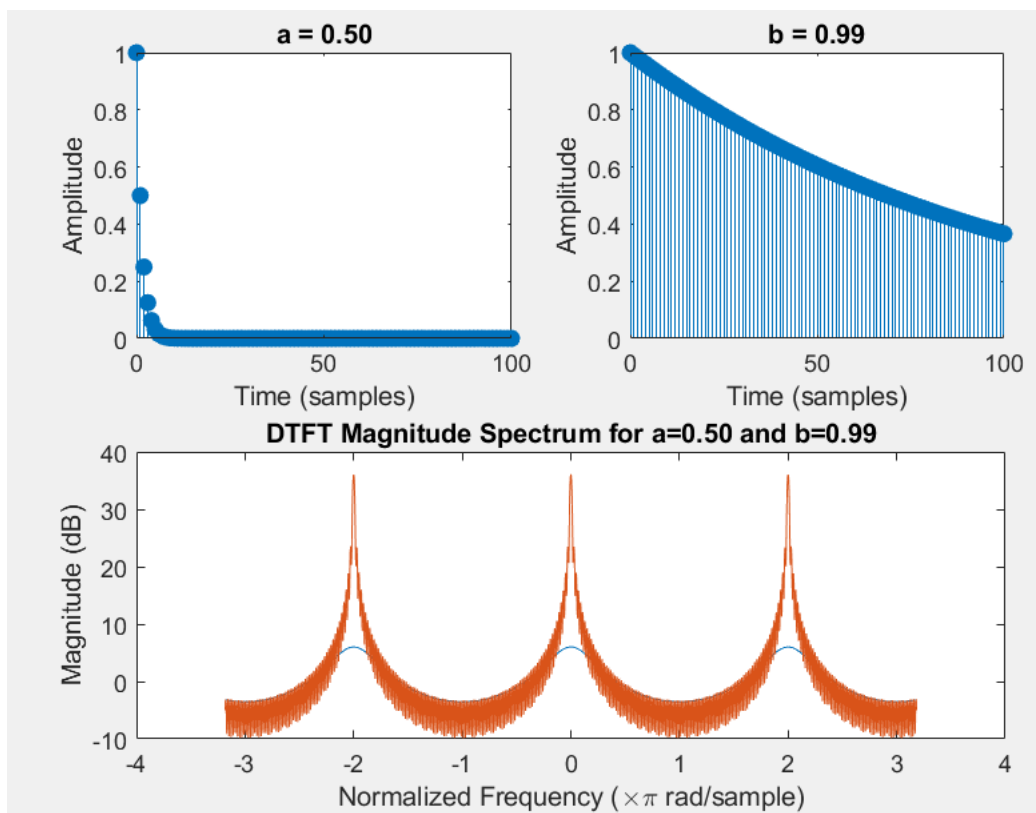


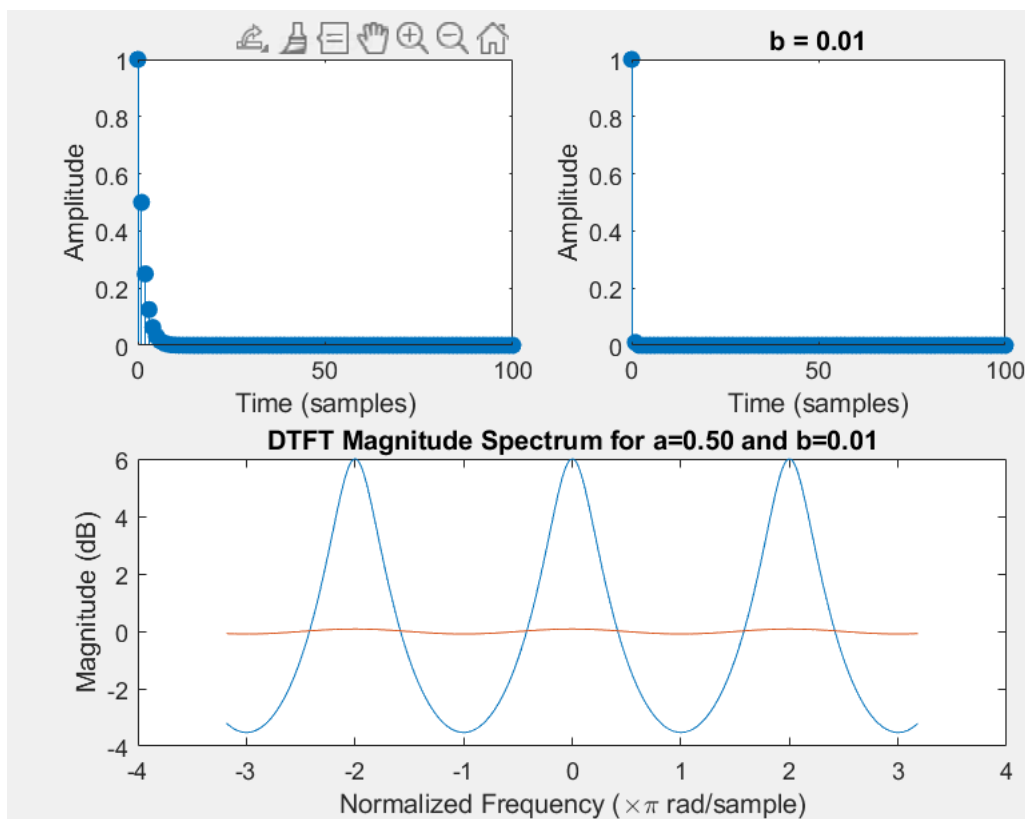
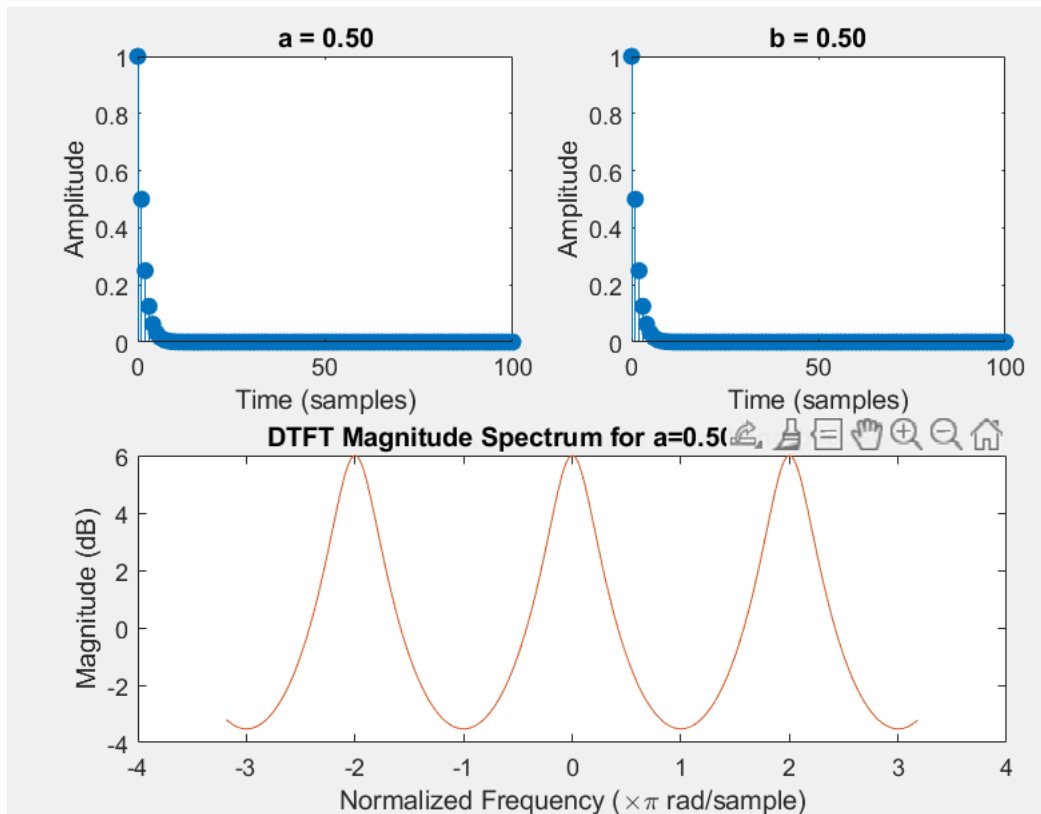
b4)

plot:



c) $\omega = -10:0.01:10$





2. Discrete-time filters:

Here we are considering the $x[n]$ as input and $y[n]$ as output.

And the relation between the input and the output is

$$y[n] = \frac{1}{M} \sum_{m=0}^{M-1} x[n-m]$$

a) here the impulse response of this system is $h[n]$ then the input $x[n]$ should be the $\delta[n]$ i.e

$$\frac{1}{M} \sum_{m=0}^{M-1} x[n-m]$$

$x[n] = \delta[n]$ then the $h[n] =$
 $m] = \delta[n-m]$

and the $x[n] = \delta[n]$ and the $x[n-m]$

The impulse response of a moving average filter with order M is a sequence that has a value of $1/M$ for the first M samples and is zero for all other samples. It can be represented as follows:

$$h[n] = (1/M) \text{ for } n = 0, 1, 2, \dots, M-1$$

$$h[n] = 0 \text{ otherwise}$$

b) $y[n] = \text{conv}(x[n], v[n])$

c) here the input signal becomes noisy when we add the $s[n]$ and $v[n]$ here the $s[n]$ is given as

$$s[n] = 5 \sin(w_0 n)$$

and the $v[n]$ is random signal.

The input signal $x[n] = s[n] + v[n]$ and we are considering the $w_0 = \pi/200$

And taking the n values from 0 to 1000

CODE:

```

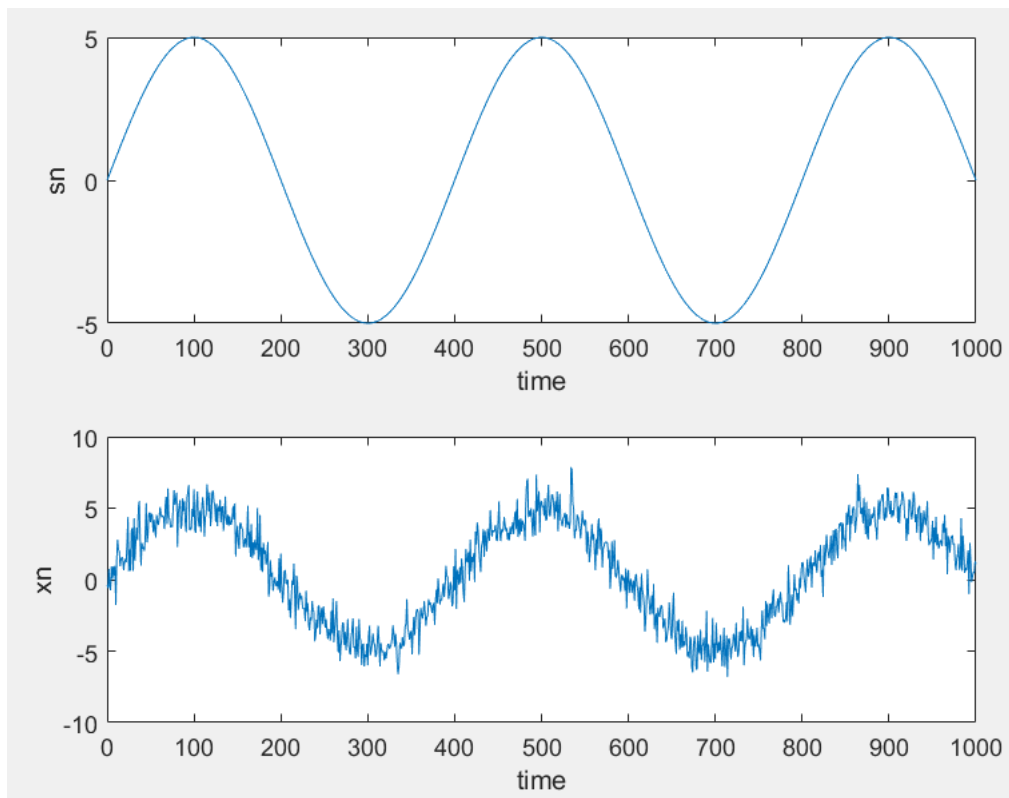
wo=pi/200;
n=0:1000;
sn=5*sin(wo*n);
vn=randn(1,1001);
xn=sn+vn;
figure
subplot(2,1,1)
plot(n,sn);
xlabel('time')
ylabel('sn')

subplot(2,1,2)
plot(n,xn)
xlabel('time')
ylabel('xn')

```

The plots of $x[n]$ and the $s[n]$ verses n are:

Plot:



d) Filter the noisy signal $x[n]$ using the moving average filter with different values of M and plot the results. Here's how you can do it:

here we are changing the M values

$M=5,21,51$

then

$$y[n] = \frac{1}{M} \sum_{m=0}^{M-1} x[n-m]$$

For finding $y[n]$ we convolute the $x[n]$ and the $h[n]$

ie $y[n]=\text{conv}(x[n],v[n])$

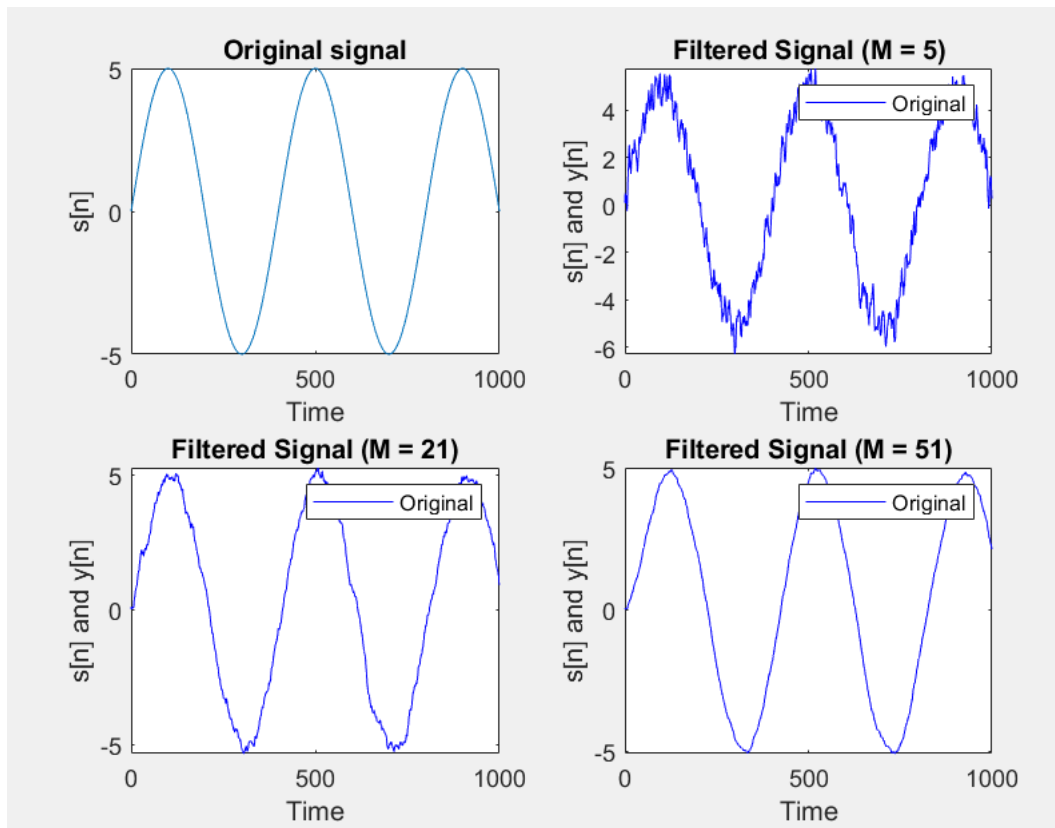
CODE:

```

1  M = [5, 21, 51];
2  n = 0:1000;
3  wn = pi / 200;
4  sn = 5 * sin(wn * n);
5  vn = randn(1, 1001);
6  xn = sn + vn;
7
8  figure;
9
10 for i = 1:length(M)
11     hn = ones(1, M(i)) / M(i);
12     yn = conv(xn, hn, 'full');
13
14     subplot(2,2,i);
15     plot(n, sn);
16     xlabel('Time');
17     ylabel('s[n]');
18     title('Original signal');
19
20     subplot(2,2,i+1);
21     plot(n, yn(1:length(n)), 'b');
22     xlabel('Time');
23     ylabel('s[n] and y[n]');
24     title(['Filtered Signal (M = ' num2str(M(i)) ')']);
25     legend('Original', 'Filtered');
26 end
27

```

Plot of $y[n]$ for all the values of M:



e)***Observation:**

Here as you change the value of M:

*smaller M values result in less smoothing and better presentation of high frequency components in the signal.

*Larger M values result in more smoothing and better noise reduction but can blur or distort the signal's sharp features.

*There's a trade-off between noise reduction and signal fidelity.

Higher M values reduce noise but may introduce distortion.

g)here the relation between the input and the output signal is given by

$$y[n] = x[n] - x[n-1]$$

For the digital differentiator filter , the impulse response $h[n]$ is given by:

$$h[n] = 1 \text{ for } n=0$$

$$h[n] = -1 \text{ for } n=1$$

$$h[n] = 0 \text{ otherwise}$$

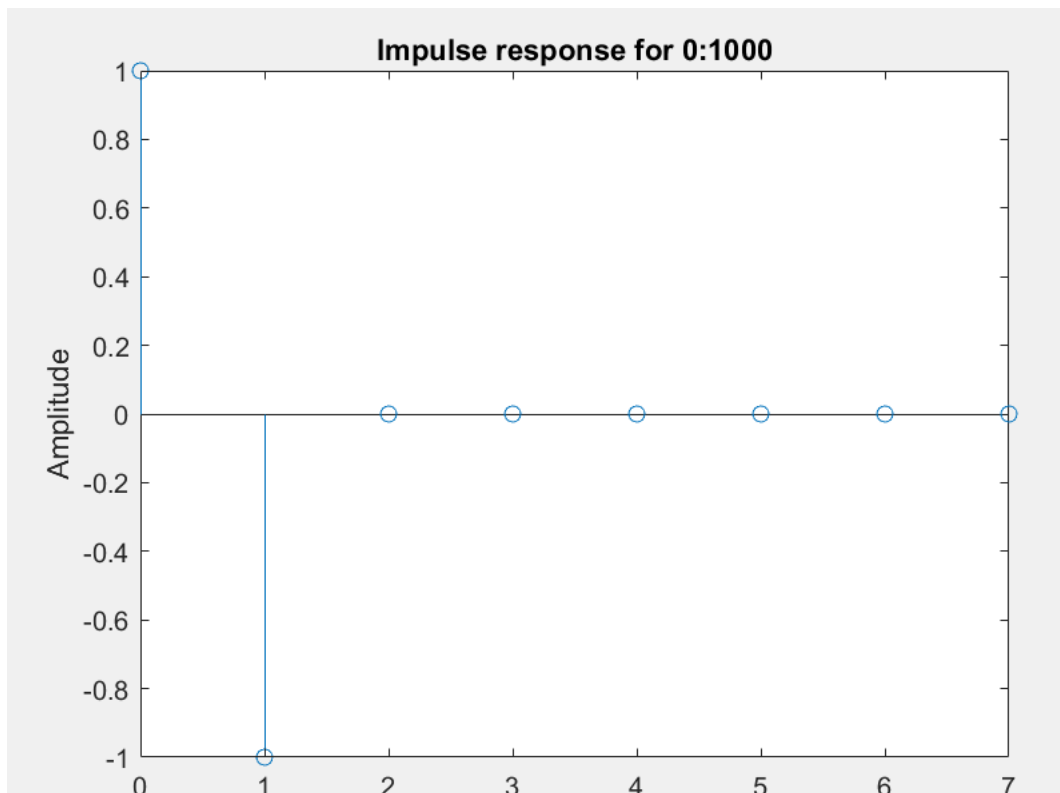
part.a)

Code:

Here n=0:1000;

```
1 sum = zeros(n);  
2 hn = [1, -1, zeros(1,6)];  
3  
4 figure;  
5 stem(0:7,hn);  
6 title(sprintf("Impulse response for %d:%d",n(1),n(end)));  
7 xlabel("Time");  
8 ylabel("Amplitude");  
9
```

Plot:



Part.c:

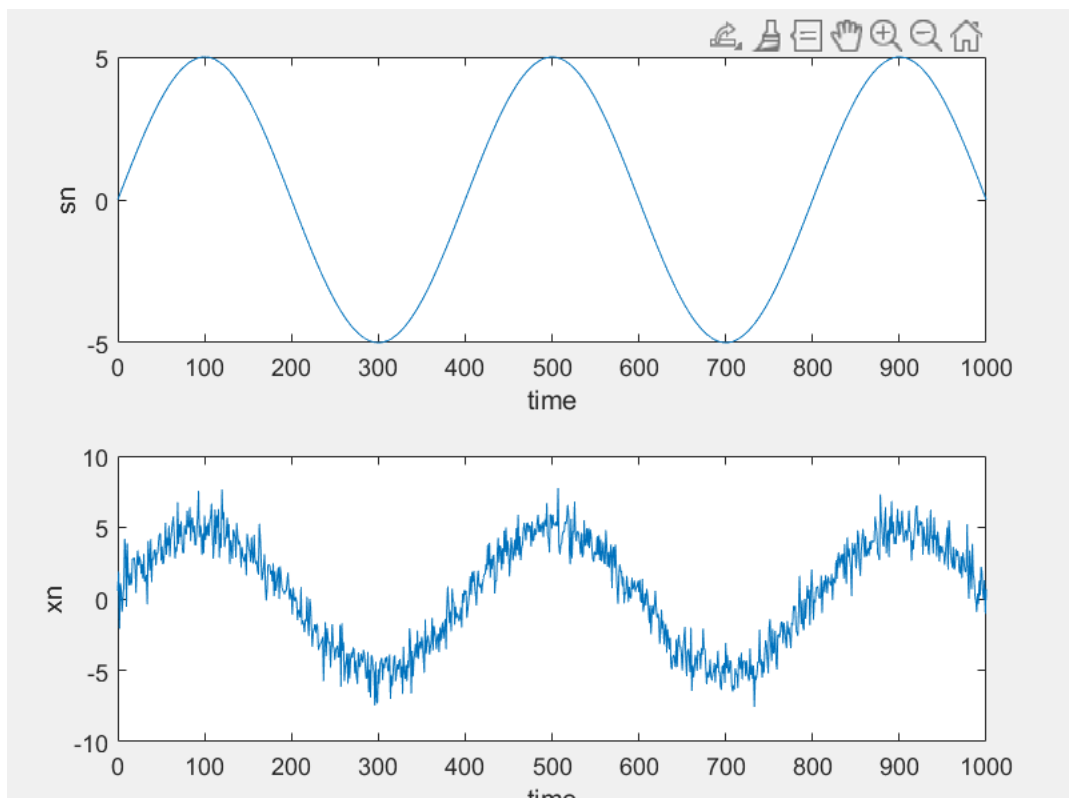
Code:

```

1  wo=pi/200;
2  n=0:1000;
3  sn=5*sin(wo*n);
4  vn=randn(1,1001);
5  xn=sn+vn;
6  figure
7  subplot(2,1,1)
8  plot(n,sn);
9  xlabel('time')
10 ylabel('sn')
11
12 subplot(2,1,2)
13 plot(n,xn)
14 xlabel('time')
15 ylabel('xn')
16

```

Plot:



Part.d)

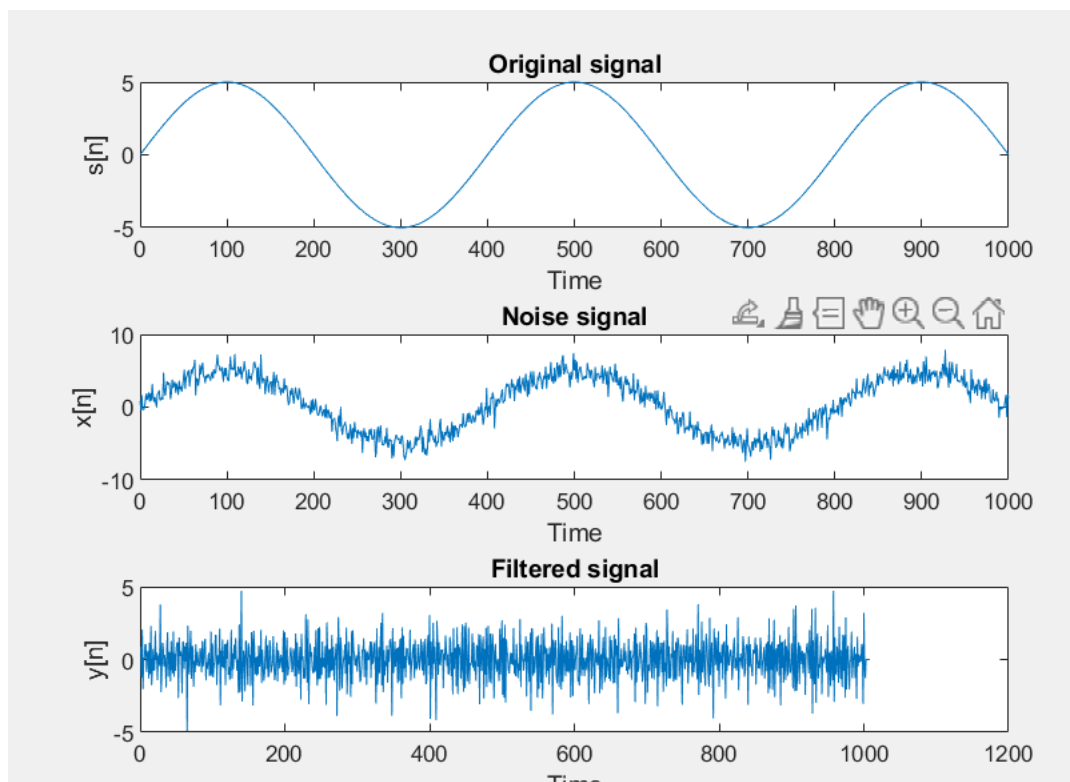
CODE:

```

1  n = 0:1000;
2  wn = pi / 200;
3  sn = 5 * sin(wn * n);
4  vn = randn(1, 1001);
5  xn = sn + vn;
6
7  figure;
8  hn = [1, -1, zeros(1,6)];
9
10     subplot(3,1,1);
11     plot(n, sn);
12     xlabel('Time');
13     ylabel('s[n]');
14     title('Original signal');
15
16
17     subplot(3,1,2);
18     plot(n, xn);
19     xlabel('Time');
20     ylabel('x[n]');
21     title('Noise signal');
22
23
24     subplot(3,1,3);
25     plot(yn);
26     xlabel('Time');
27     ylabel('y[n]');
28     title('Filtered signal');
29
30

```

Plot:



Part.f

CODE:

```
4     N = 1000;
5     omega = pi / 200;
6     n = 0:N;
7     s = 5 * sin(omega * n);
8     w = -10: 0.01: 10;
9     v = randn(size(n));
10    x = s + v;
11
12
13    t = 10;
14    filter = [1, -1, zeros(1, t-2)];
15
16    y = conv(x, filter, 'full');
17
18    X_noisy = DT_Fourier(x, 1, w);
19    Y_filtered = DT_Fourier(y, 1, w);
20
21    figure(1)
22    subplot(2, 2, 1);
23    plot(w, abs(X_noisy));
24    title('Magnitude Spectrum of Noisy Signal');
25    subplot(2, 2, 2);
26    plot(w, angle(X_noisy));
27    title('Phase Spectrum');
28    subplot(2, 2, 3);
29    plot(w, real(X_noisy));
30    title('Real Part Spectrum');
31    subplot(2, 2, 4);
32    plot(w, imag(X_noisy));
33    title('Imaginary Part Spectrum');
34
35    figure(2)
36    subplot(2, 2, 1);
37    plot(w, abs(Y_filtered));
38    title('Magnitude Spectrum of Filtered Signal');
39    subplot(2, 2, 2);
40    plot(w, angle(Y_filtered));
41    title('Phase Spectrum');
42    subplot(2, 2, 3);
43    plot(w, real(Y_filtered));
44    title('Real Part Spectrum');
45    subplot(2, 2, 4);
46    plot(w, imag(Y_filtered));
```

h)

Frequency Selectivity:

*The moving average filter is a low-pass filter that attenuates high-frequency components while preserving low-frequency components . Its frequency selectivity is determined by the filter order M.

Smaller values of M allow higher frequencies to pass through , while larger values of M result in a more significant attenuation of high-frequency components.

EFFECT:

It is effective for smoothing signals and reducing noise but may blur or distort signals with rapid changes.

*Digital Differentiator Filter:

The digital differentiator filter enhances high-frequency components and attenuates low-frequency components. It acts as a high-pass filter. It approximates the first derivative of the signal.

EFFECT:

It can emphasize edges and rapid changes in the signal, making it useful for detecting abrupt transitions or fine details.

** The nature of the two filters is as follows:

The moving average filter is low-pass and attenuates high frequencies, while the digital differentiator filter is high-pass and enhances high-frequency information. The choice of filter depends on the specific signal processing requirements and the desired trade-off between noise reduction and feature preservation.

3)

Here the inverse DTFT is calculated by using the expression

$$x[n] = \frac{1}{2\pi} \int_{-\pi}^{\pi} X(e^{j\omega}) e^{j\omega n} d\omega.$$

Here the $X(e^{j\omega})$ is given as

$$X(e^{j\omega}) = \begin{cases} 1, & |\omega| \leq \omega_c \\ 0, & \omega_c < |\omega| < \pi \end{cases}$$

a)

here

$\omega_c = \pi/16$

and $n = -100:100$

we are plotting the real and complex valued of $x[n]$ in the interval $[-\pi, \pi]$

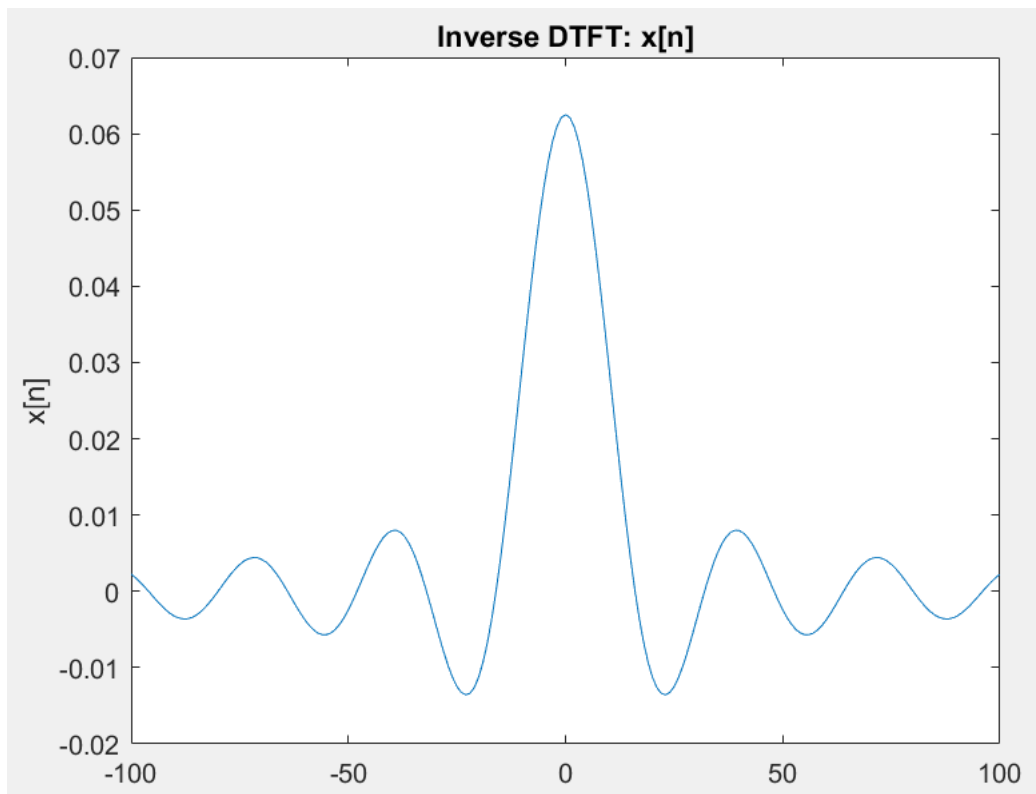
CODE:

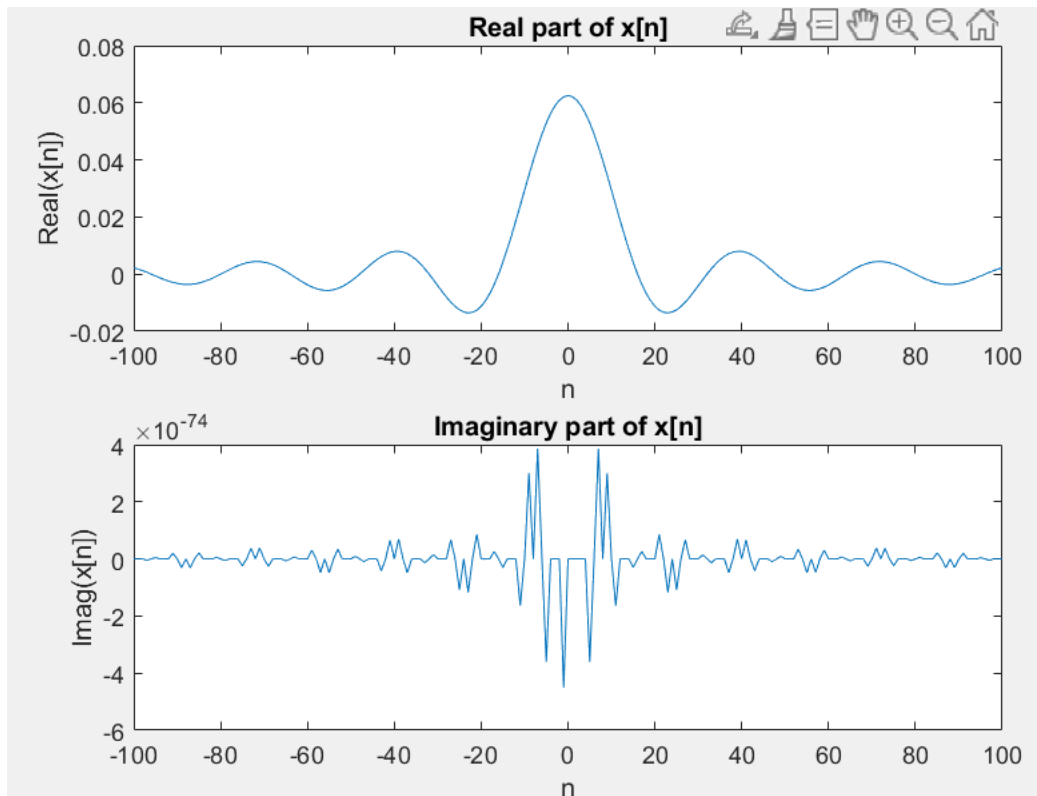
```

1  wc = pi/16;
2  n = -100:100;
3  x = zeros(1, length(n));
4  syms w;
5  X = piecewise(abs(w) <= wc, 1, wc < abs(w) & abs(w) < pi, 0);
6
7  for k = 1:length(n)
8      t = exp(1j * w * n(k));
9      x(k) = (1/(2*pi)) * int(X * t , w, -pi, pi);
10 end
11
12 figure;
13 subplot(2,1,1);
14 plot(n, real(x));
15 title('Real part of x[n]');
16 xlabel('n');
17 ylabel('Real(x[n])');
18
19 subplot(2,1,2);
20 plot(n, imag(x));
21 title('Imaginary part of x[n]');
22 xlabel('n');
23 ylabel('Imag(x[n])');
24
25 figure;
26 plot(n, real(x));
27 title('Inverse DTFT: x[n]');
28 xlabel('n');
29 ylabel('x[n]');
30

```

Plot:





Here the $x[n]$ is both real and imaginary those can be seen in the plots above clearly.

b) By changing the w_c , you are altering the bandwidth of the rectangular frequency domain signal.

*As w_c increases the rectangular frequency domain signal becomes narrower in frequency, resulting in a broader time domain signal and vice-versa

*when $w_c = \pi$ the entire frequency domain signal becomes zero (no signal), which means $x[n]$ will also be zero for all values of n . This is because the signal bandwidth has become zero, resulting in no signal in the time domain.

CODE:

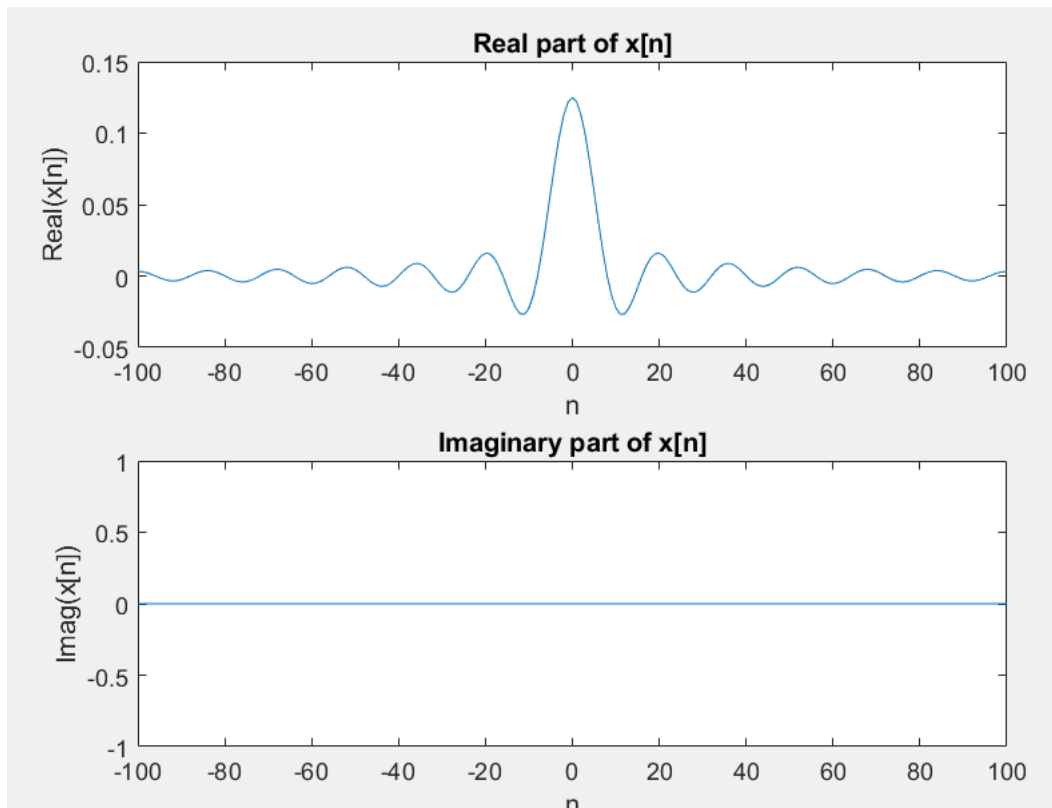

```

1  wc = pi;
2  n = -100:100;
3  x = zeros(1, length(n));
4  syms w;
5  X = piecewise(abs(w) <= wc, 1, wc < abs(w) & abs(w) < pi, 0);
6
7  for k = 1:length(n)
8      t = exp(1j * w * n(k));
9      x(k) = (1/(2*pi)) * int(X * t , w, -pi, pi);
10 end
11
12 figure;
13 subplot(2,1,1);
14 plot(n, real(x));
15 title('Real part of x[n]');
16 xlabel('n');
17 ylabel('Real(x[n])');
18
19 subplot(2,1,2);
20 plot(n, imag(x));
21 title('Imaginary part of x[n]');
22 xlabel('n');
23 ylabel('Imag(x[n])');
24
25 figure;
26 plot(n, real(x));
27 title('Inverse DTFT: x[n]');
28 xlabel('n');
29 ylabel('x[n]');
30

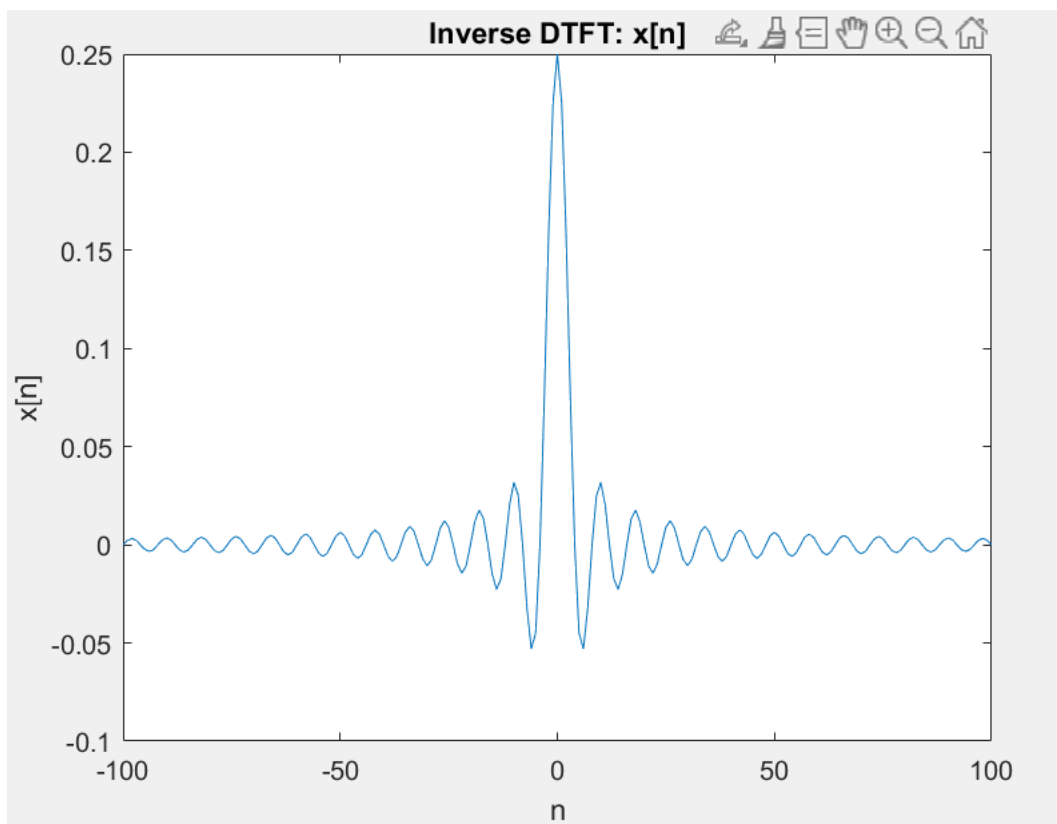
```

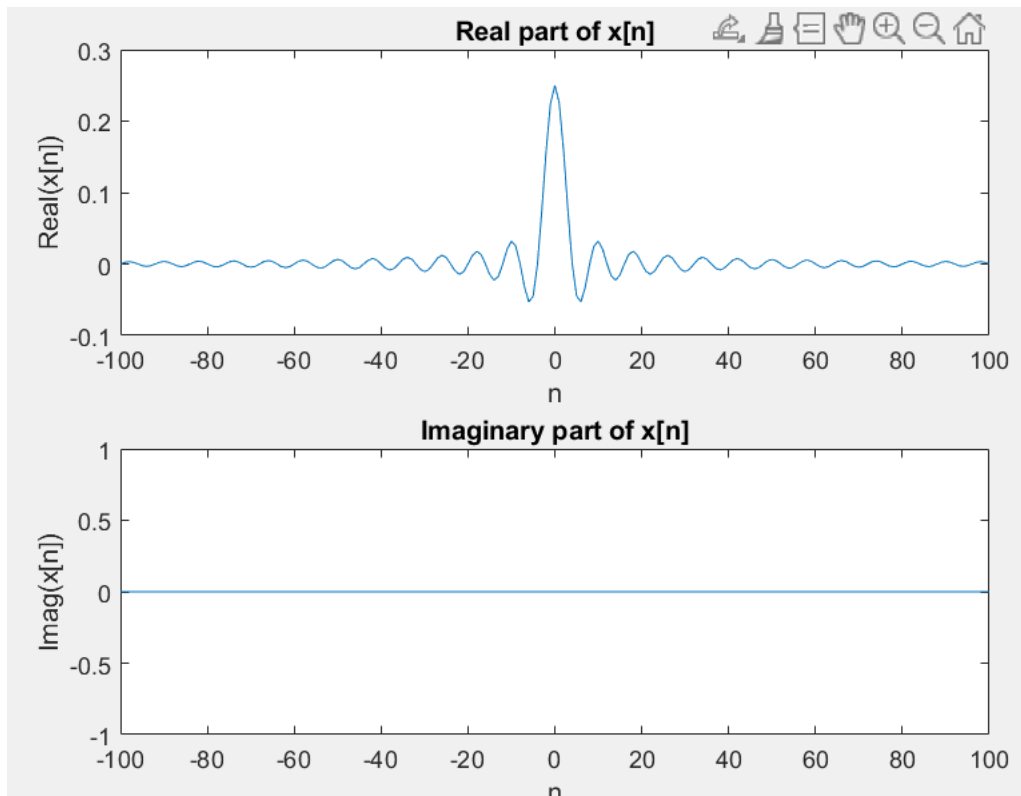
Plot:

a. for $w_c = \pi/8$

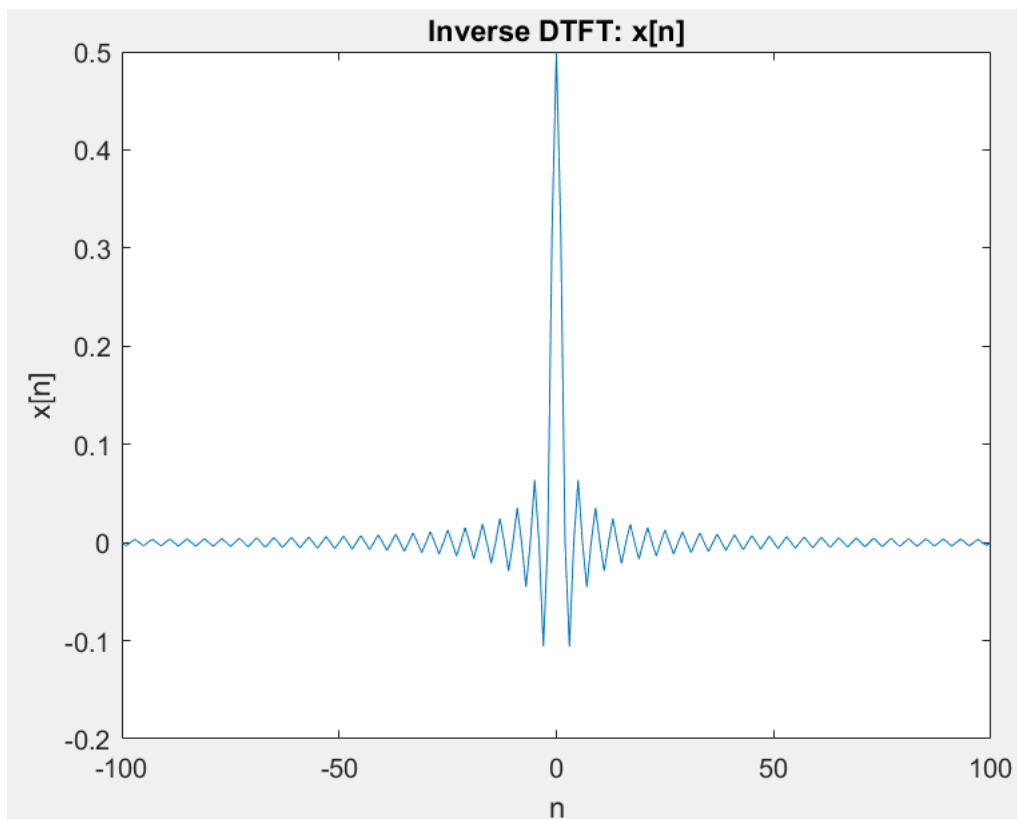


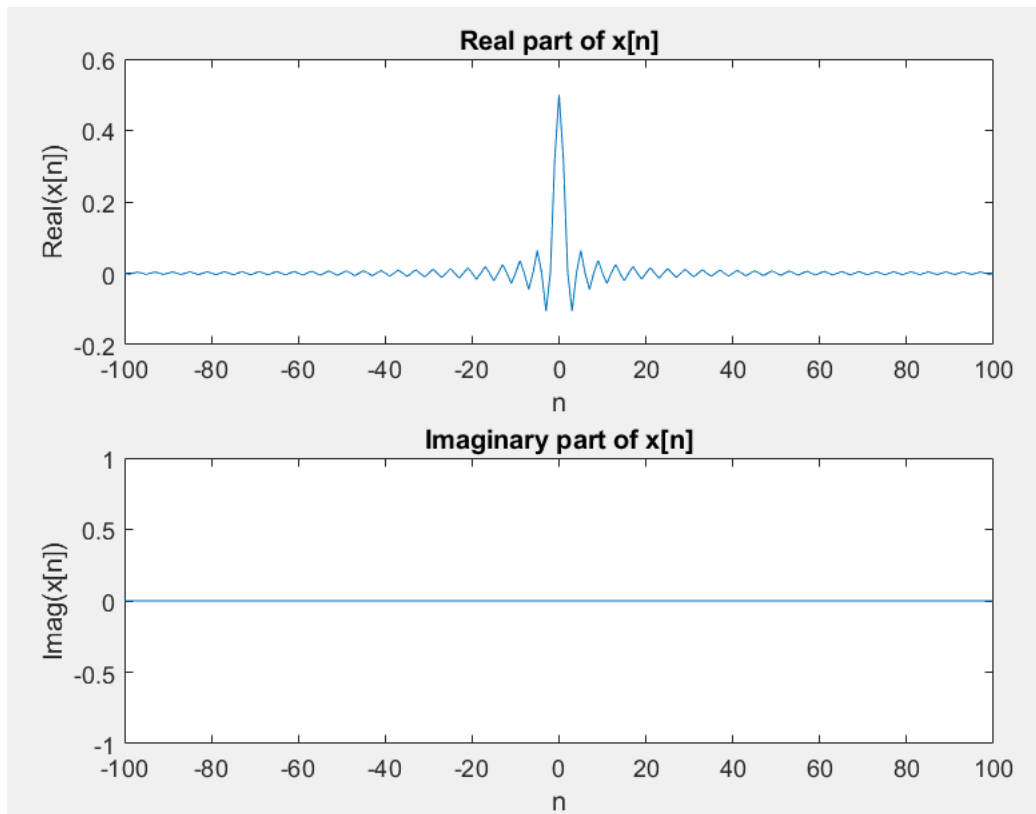
For $\omega_c = \pi/4$



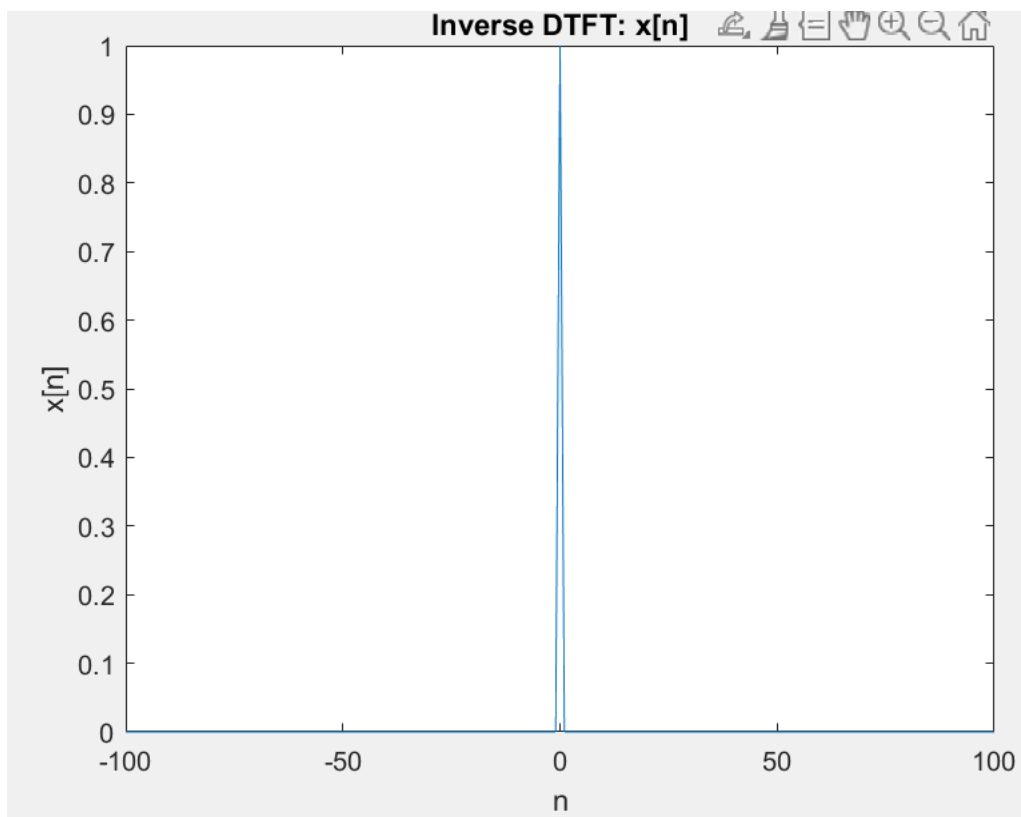


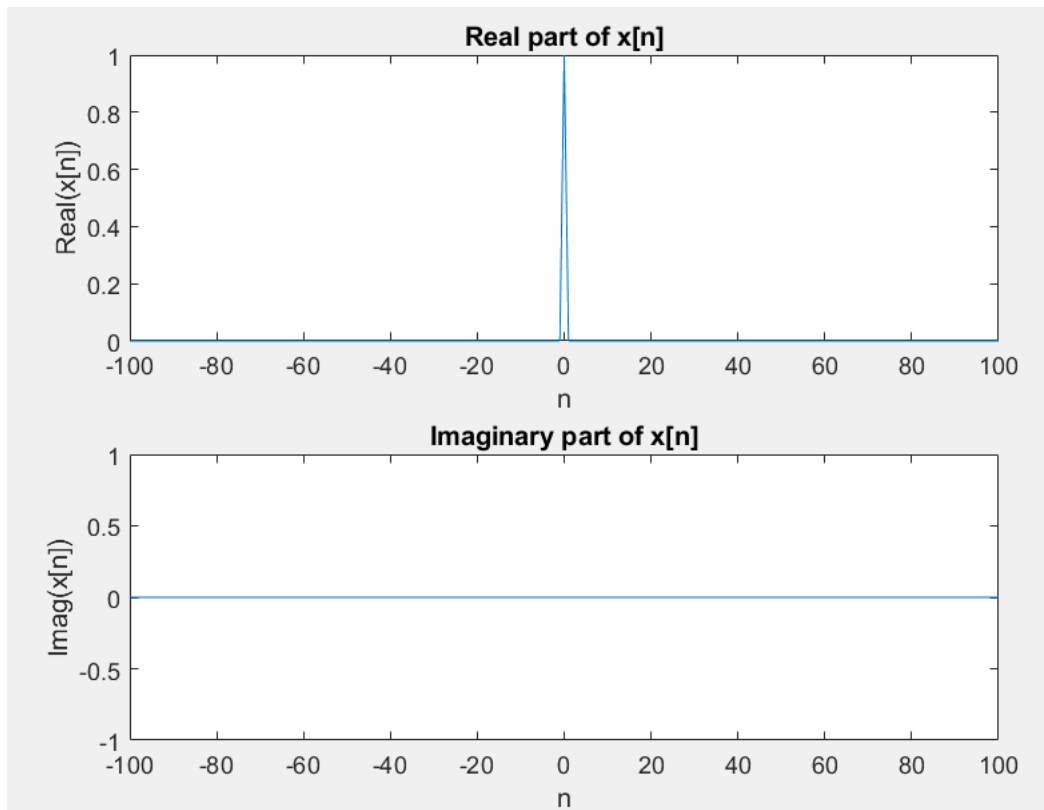
For $\omega_c = \pi/2$





For $\omega_c = \pi$





c)

here the $w_1 = \pi/8$

$w_2 = \pi/4$

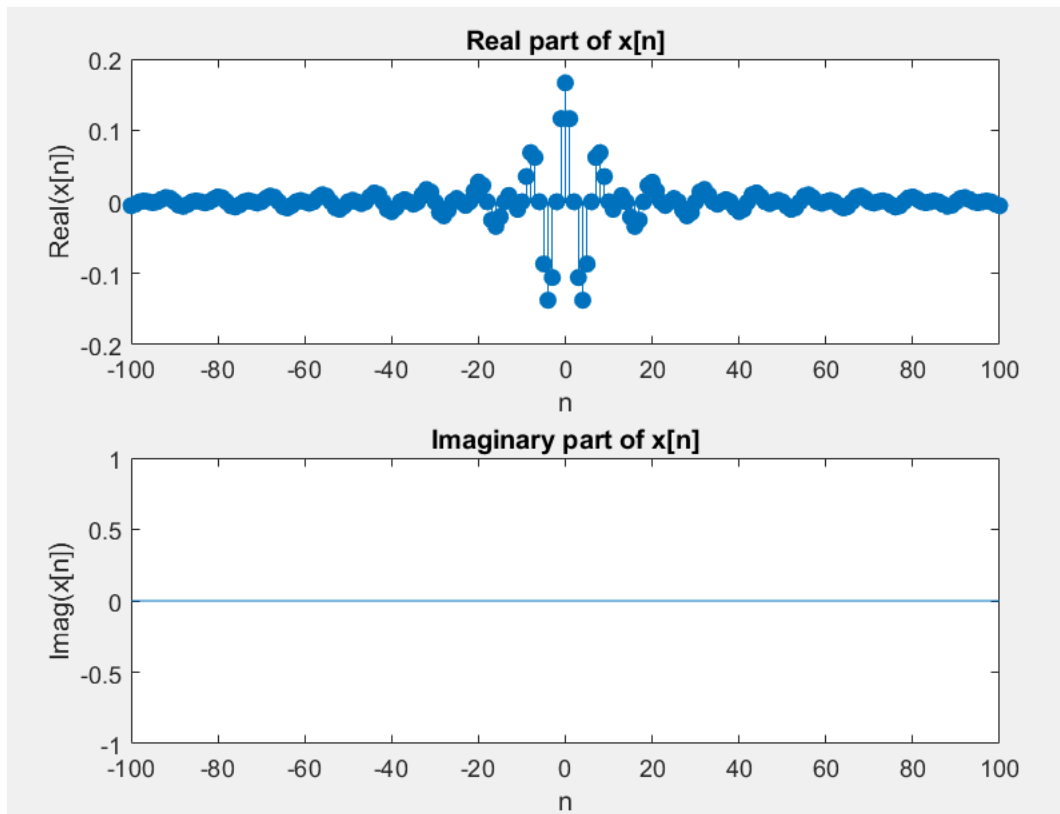
CODE:

```

1  n = -100:100;
2  x = zeros(1, length(n));
3  w1=pi/8;
4  w2=pi/4;
5
6  syms w;
7  assume(w,'real');
8  X(w)= piecewise(w1<=abs(w) & abs(w)<=w2,1,0);
9
10 for k = 1:length(n)
11     t = exp(1j * w * n(k));
12     x(k) = (1/(2*pi)) * int(X(w) * t , w, -pi, pi);
13 end
14
15 figure;
16 subplot(2,1,1);
17 plot(n, real(x));
18 title('Real part of x[n]');
19 xlabel('n');
20 ylabel('Real(x[n])');
21
22 subplot(2,1,2);
23 plot(n, imag(x));
24 title('Imaginary part of x[n]');
25 xlabel('n');
26 ylabel('Imag(x[n])');
27
28 figure;
29 plot(n, real(x));
30 title('Inverse DTFT: x[n]');
31 xlabel('n');
32 ylabel('x[n]');

```

Plot:



When the $w_1=\pi/6$ $w_2=\pi/3$

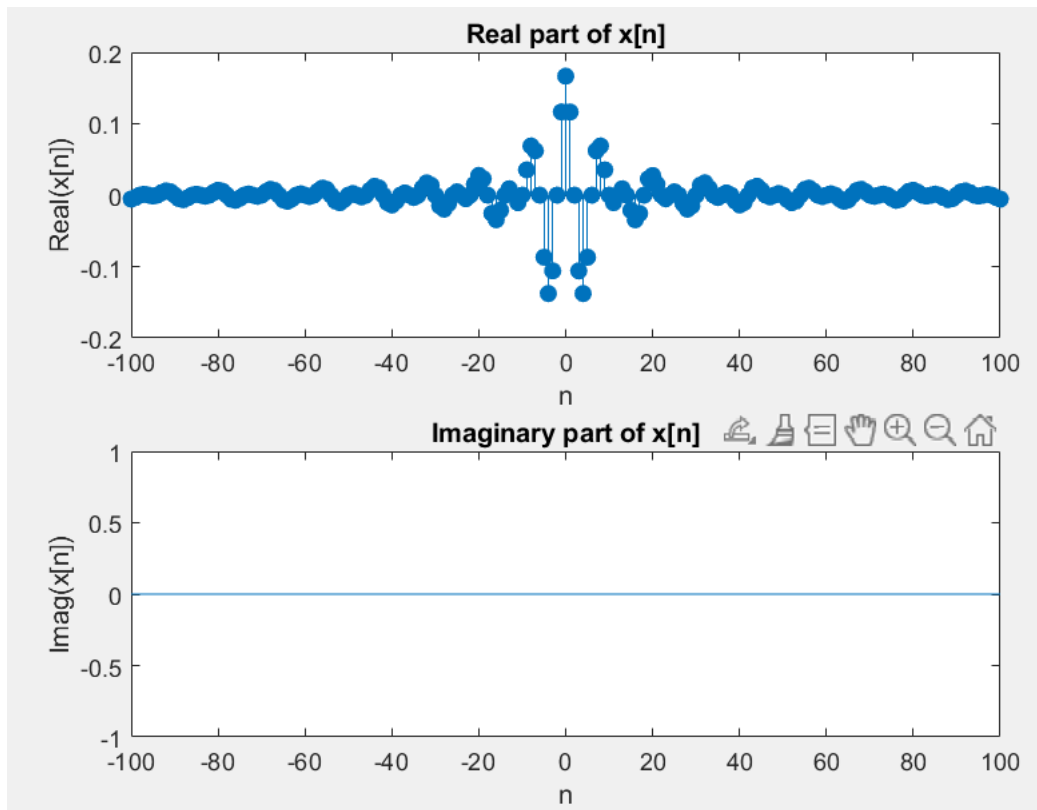
Code:

```

1  n = -100:100;
2  x = zeros(1, length(n));
3  w1=pi/6;
4  w2=pi/3;
5
6  syms w;
7  assume(w, 'real');
8  X= piecewise(w1<=abs(w) & abs(w)<=w2,1,0);
9
10 for k = 1:length(n)
11     t = exp(1j * w * n(k));
12     x(k) = (1/(2*pi)) * int(X * t , w, -pi, pi);
13 end
14
15 figure;
16 subplot(2,1,1);
17 stem(n, real(x), "filled");
18 title('Real part of x[n]');
19 xlabel('n');
20 ylabel('Real(x[n])');
21
22 subplot(2,1,2);
23 plot(n, imag(x));
24 title('Imaginary part of x[n]');
25 xlabel('n');
26 ylabel('Imag(x[n])');
27
28 figure;
29 plot(n, real(x));
30 title('Inverse DTFT: x[n]');
31 xlabel('n');
32 ylabel('x[n]');

```

Plots:



A band-pass signal with different values of w_1 and w_2 are considered .

*this part explores band-pass signals , where w_1 and w_2 define the frequency range of the signal.

*For the given values of w_1 and w_2 , the script computes $x[n]$ and plots it for each case.

By changing w_1 , w_2 you can observe how the bandwidth of the signal affects the time-domain signal $x[n]$. Narrower bands result in longer duration signals , and wider bands result in shorter duration signals.

This explains the frequency content of a signal in the frequency domain relates to its time domain representation through the inverse DTFT . different values of parameters like w_1, w_c, w_2 lead to different time domain signals , explains relationship between frequency and time domains.