

Assignment No 9

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1 Aim

In this assignment, we continue our analysis of signals using Fourier Transforms. This time, we focus on finding transforms of non periodic functions.

2 Examples

The worked examples in the assignment are given below: Spectrum of $\sin(\sqrt{2}t)$ is given below

```
t=linspace(-pi,pi,65);t=t[: -1]
dt=t[1]-t[0];fmax=1/dt
y=sin(sqrt(2)*t)
y[0]=0 # the sample corresponding to -tmax should be set zero
y=fftshift(y) # make y start with y(t=0)
Y=fftshift(fft(y))/64.0
w=linspace(-pi*fmax,pi*fmax,65);w=w[: -1]
figure()
subplot(2,1,1)
plot(w,abs(Y),lw=2)
xlim([-10,10])
ylabel(r"$|Y|$",size=16)
title(r"Spectrum of $\sin\left(\sqrt{2}t\right)$")
grid(True)
subplot(2,1,2)
plot(w,angle(Y),'ro',lw=2)
xlim([-10,10])
ylabel(r"Phase of $Y$",size=16)
xlabel(r"$\omega$",size=16)
grid(True)
```

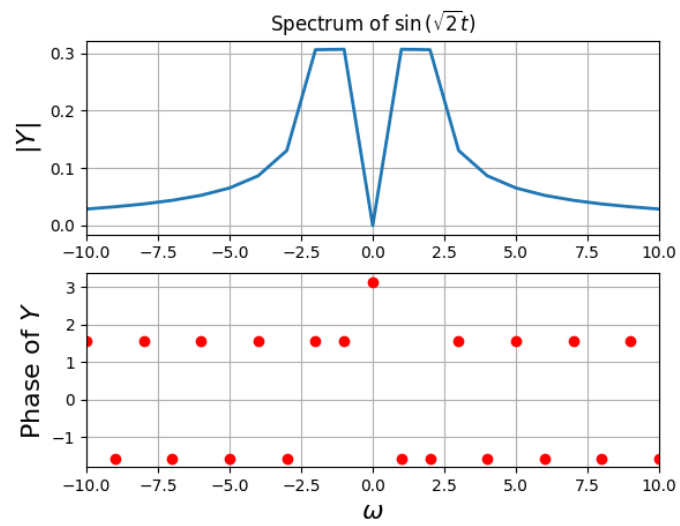


Figure 1: Spectrum of $\sin(\sqrt{2}t)$

Original function for which we want the DFT:

```
t1=linspace(-pi,pi,65);t1=t1[: -1]
t2=linspace(-3*pi,-pi,65);t2=t2[: -1]
t3=linspace(pi,3*pi,65);t3=t3[: -1]
# y=sin(sqrt(2)*t)
figure()
plot(t1,sin(sqrt(2)*t1),'b',lw=2)
plot(t2,sin(sqrt(2)*t2),'r',lw=2)
plot(t3,sin(sqrt(2)*t3),'r',lw=2)
ylabel(r"$y$",size=16)
xlabel(r"$t$",size=16)
title(r"$\sin\left(\sqrt{2}t\right)$")
grid(True)
```

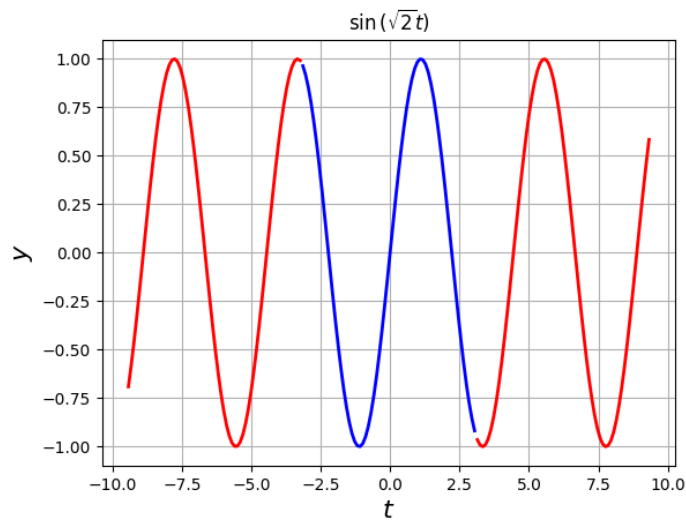


Figure 2: $\sin(\sqrt{2}t)$

As the DFT is computed over a finite time interval, we have actually plotted the DFT for this function

```
t1=linspace(-pi,pi,65);t1=t1[: -1]
t2=linspace(-3*pi,-pi,65);t2=t2[: -1]
t3=linspace(pi,3*pi,65);t3=t3[: -1]
y=sin(sqrt(2)*t1)
figure()
plot(t1,y,'bo',lw=2)
plot(t2,y,'ro',lw=2)
plot(t3,y,'ro',lw=2)
ylabel(r"$y$",size=16)
```

```

xlabel(r"$t$",size=16)
title(r"$\sin\left(\sqrt{2}t\right)$ with $t$ wrapping every $2\pi$ ")
grid(True)

```

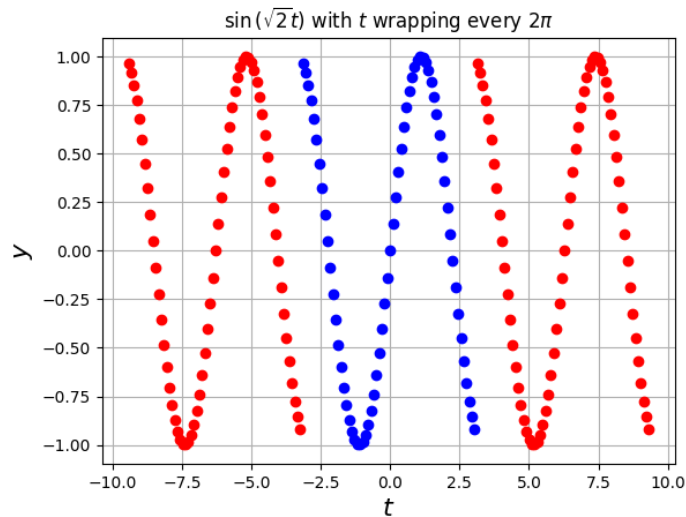


Figure 3: Spectrum of $\sin(\sqrt{2}t)$

These discontinuities lead to non harmonic components in the FFT which decay as $\frac{1}{\omega}$. To confirm this, we plot the spectrum of the periodic ramp.

```

t=linspace(-pi,pi,65);t=t[: -1]
dt=t[1]-t[0];fmax=1/dt
y=t
y[0]=0 # the sample corresponding to -tmax should be set zeroo
y=fftshift(y) # make y start with y(t=0)
Y=fftshift(fft(y))/64.0
w=linspace(-pi*fmax,pi*fmax,65);w=w[: -1]
figure()
semilogx(abs(w),20*log10(abs(Y)),lw=2)
xlim([1,10])
ylim([-20,0])
xticks([1,2,5,10],["1","2","5","10"],size=16)
ylabel(r"$|Y|$ (dB)",size=16)
title(r"Spectrum of a digital ramp")
xlabel(r"$\omega$",size=16)
grid(True)

```

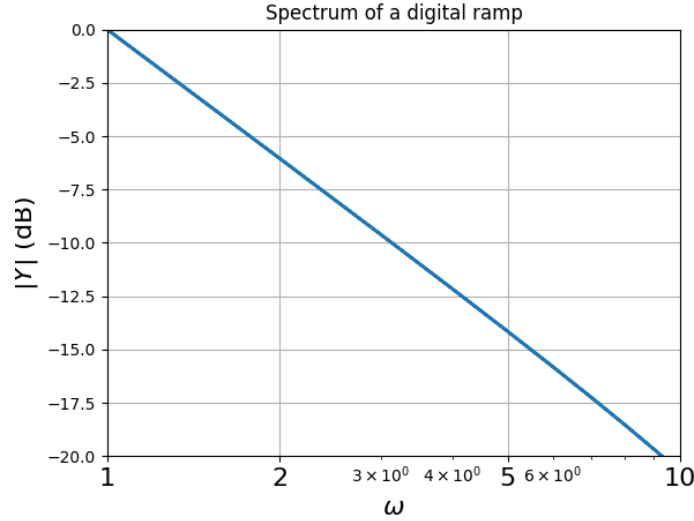


Figure 4: Spectrum of $\sin(\sqrt{2}t)$

2.1 Windowing

The hamming window removes discontinuities by attenuating the high frequency components that cause the discontinuities. The hamming window function is given by

$$x[n] = 0.54 + 0.46\cos\left(\frac{2\pi n}{N-1}\right) \quad (1)$$

We multiply our signal with the hamming window and periodically extend it. The discontinuities nearly vanish.

```
t1=linspace(-pi,pi,65);t1=t1[:-1]
t2=linspace(-3*pi,-pi,65);t2=t2[:-1]
t3=linspace(pi,3*pi,65);t3=t3[:-1]
n=arange(64)
wnd=fftshift(0.54+0.46*cos(2*pi*n/63))
y=sin(sqrt(2)*t1)*wnd
figure()
plot(t1,y,'bo',lw=2)
plot(t2,y,'ro',lw=2)
plot(t3,y,'ro',lw=2)
ylabel(r"$y$",size=16)
xlabel(r"$t$",size=16)
title(r"$\sin\left(\sqrt{2}t\right)\times w(t)$ with $t$ wrapping every $")
grid(True)
```

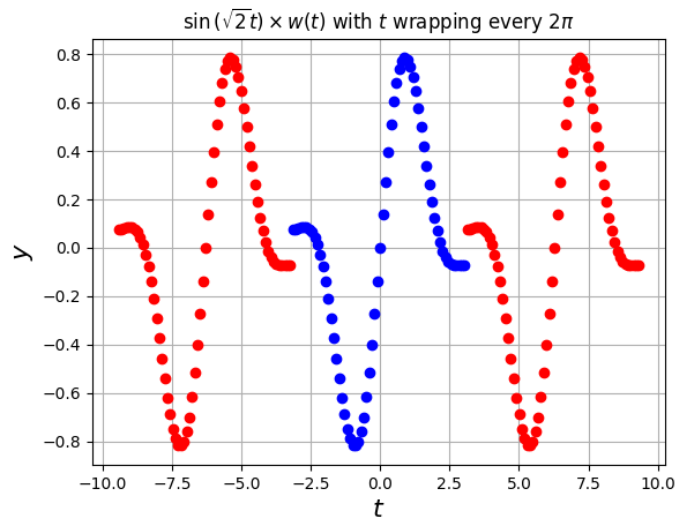


Figure 5: Spectrum of $\sin(\sqrt{2}t) * w(t)$

The spectrum that is obtained with a time period 2π is given below:

```
t=linspace(-pi,pi,65);t=t[: -1]
dt=t[1]-t[0];fmax=1/dt
n=arange(64)
wnd=fftshift(0.54+0.46*cos(2*pi*n/63))
y=sin(sqrt(2)*t)*wnd
y[0]=0 # the sample corresponding to -tmax should be set zero
y=fftshift(y) # make y start with y(t=0)
Y=fftshift(fft(y))/64.0
w=linspace(-pi*fmax,pi*fmax,65);w=w[: -1]
figure()
subplot(2,1,1)
plot(w,abs(Y),lw=2)
xlim([-8,8])
ylabel(r"$|Y|$",size=16)
title(r"Spectrum of $\sin\left(\sqrt{2}t\right)\times w(t)$")
grid(True)
subplot(2,1,2)
plot(w,angle(Y),'ro',lw=2)
xlim([-8,8])
ylabel(r"Phase of Y",size=16)
xlabel(r"$\omega$",size=16)
grid(True)
```

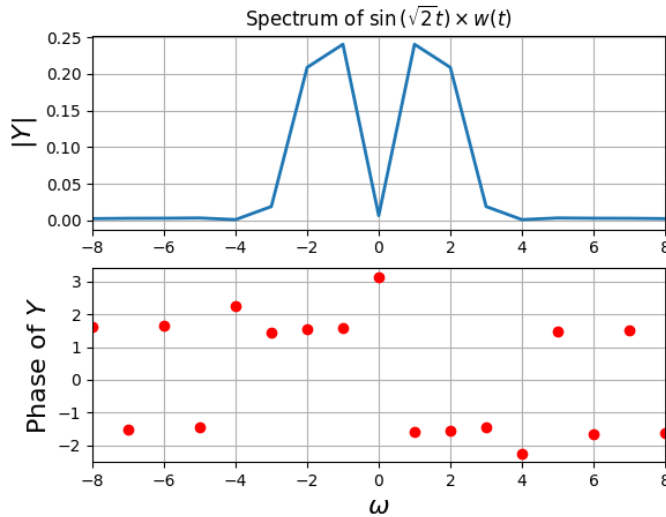


Figure 6: Spectrum of $\sin(\sqrt{2}t) * w(t)$

The spectrum that is obtained with a time period 8π has a sharper peak

and is given below:

```

t=linspace(-4*pi,4*pi,257);t=t[:-1]
dt=t[1]-t[0];fmax=1/dt
n=arange(256)
wnd=fftshift(0.54+0.46*cos(2*pi*n/256))
y=sin(sqrt(2)*t)
# y=sin(1.25*t)
y=y*wnd
y[0]=0 # the sample corresponding to -tmax should be set zeroo
y=fftshift(y) # make y start with y(t=0)
Y=fftshift(fft(y))/256.0
w=linspace(-pi*fmax,pi*fmax,257);w=w[:-1]
figure()
subplot(2,1,1)
plot(w,abs(Y),'b',w,abs(Y),'bo',lw=2)
xlim([-4,4])
ylabel(r"$|Y|$",size=16)
title(r"Spectrum of $\sin\left(\sqrt{2}t\right)\times w(t)$")
grid(True)
subplot(2,1,2)
plot(w,angle(Y),'ro',lw=2)
xlim([-4,4])
ylabel(r"Phase of $Y$",size=16)
xlabel(r"$\omega$",size=16)
grid(True)

```

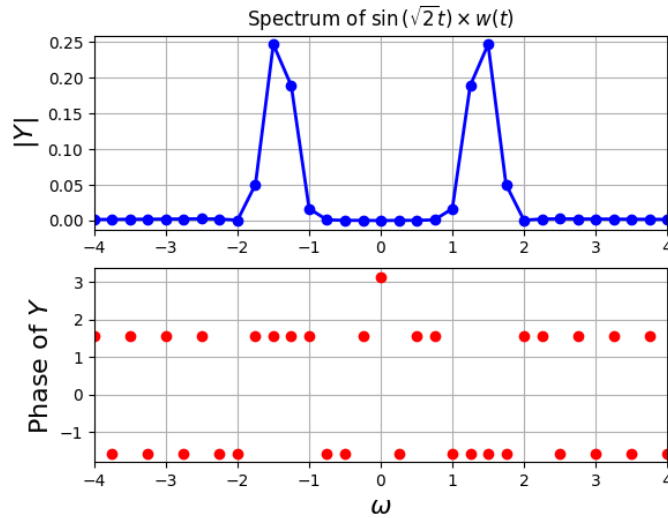



Figure 7: Spectrum of $\sin(\sqrt{2}t) * w(t)$

3 Questions

3.1 Question 2

In this question, we shall plot the FFT of $\cos^3(0.86t)$ The FFT without the hamming Window:

```

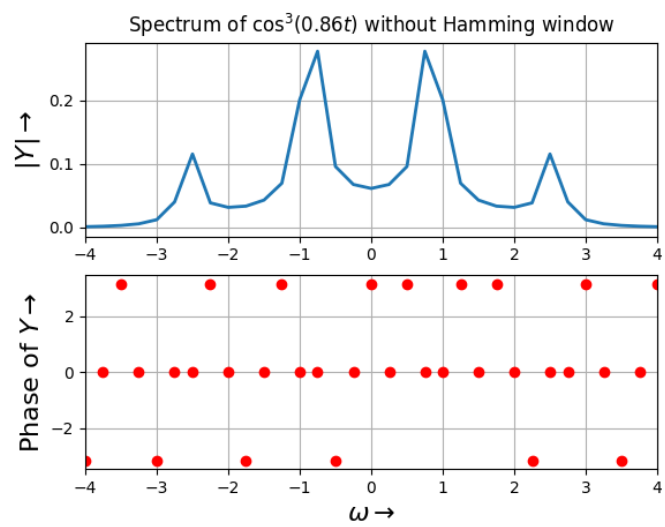
y = cos(0.86*t)**3
yw = y*wnd
y[0]=0
yw[0]=0
y = fftshift(y)
yw = fftshift(yw)
Y = fftshift(fft(y))/256.0           #without hamming window
Yw = fftshift(fft(yw))/256.0        #with hamming window

```

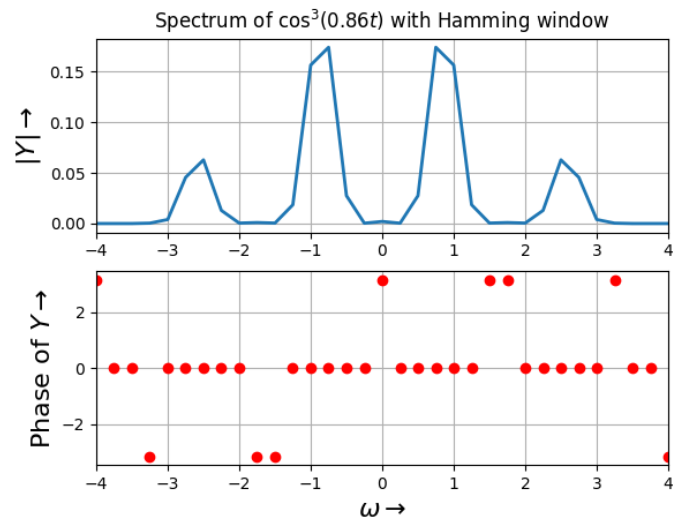
```

plot_spectrum(2,w,Y,4,r" Spectrum of  $\cos^3(0.86t)$  without Hamming window")
plot_spectrum(3,w,Yw,4,r" Spectrum of  $\cos^3(0.86t)$  with Hamming window")

```



The FFT with the hamming Window:



We notice that a lot of the energy is stored in frequencies that aren't a part of the signal. After windowing, these frequencies are attenuated and hence the peaks are sharper in the windowed function.

3.2 Question 3

We need to estimate ω and δ for a signal $\cos(\omega t + \delta)$ for 128 samples between $[-\pi, \pi)$. We estimate ω using a weighted average. We have to extract the digital spectrum of the signal and find the two peaks at $\pm\omega_0$, and estimate ω and δ .

```
w0 = 0.8
d = 0.5
t = linspace(-pi, pi, 129)[: -1]
dt = t[1] - t[0]; fmax = 1/dt
n = arange(128)
wnd = fftshift(0.54 + 0.46 * cos(2 * pi * n / 128))
y = cos(w0 * t + d) * wnd
y[0] = 0
y = fftshift(y)
Y = fftshift(fft(y)) / 128.0
w = linspace(-pi * fmax, pi * fmax, 129); w = w[: -1]
plot_spectrum(4, w, Y, 4, r"Digital Spectrum of $\cos(w_0t + \delta)$", r"$|Y|$")

ii = where(w >= 0)
cal_w = sum(abs(Y[ii]) ** 2 * w[ii]) / sum(abs(Y[ii]) ** 2)
i = abs(w - cal_w).argmin()
delta = angle(Y[i])
print("Value of w0 without noise: ", cal_w)
print("Value of delta without noise: ", delta)
```

We estimate ω by performing a Mean average of ω over the magnitude of $|Y(j\omega)|$. For δ we consider a window on each half of ω (split into positive and negative values) and extract their mean slope.

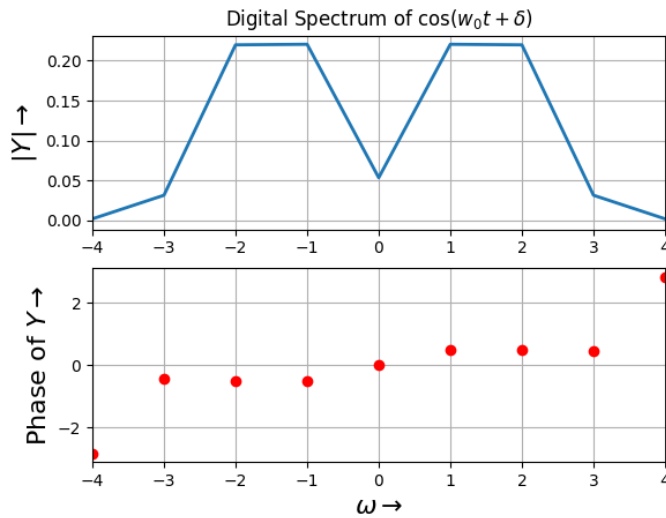


Figure 8: Fourier transform of $\cos(1.5t + 0.5)$

3.3 Question 4

We repeat the exact same process as question 3 but with noise added to the original signal.

```

y = (cos(w0*t + d) + 0.1*randn(128))*wnd
y[0]=0
y = fftshift(y)
Y = fftshift(fft(y))/128.0
plot_spectrum(5,w,Y,4,r" Spectrum of a noisy  $\cos(w_0 t + \delta)$  with Hamr

# w0 is calculated by finding the weighted average of all w>0. Delta is f
ii = where(w>=0)
w_cal = sum(abs(Y[ii])**2*w[ii])/sum(abs(Y[ii])**2)
i = abs(w-w_cal).argmin()
delta = angle(Y[i])
print("Calculated value of w0 with noise: ",w_cal)
print("Calculated value of delta with noise: ",delta)

```

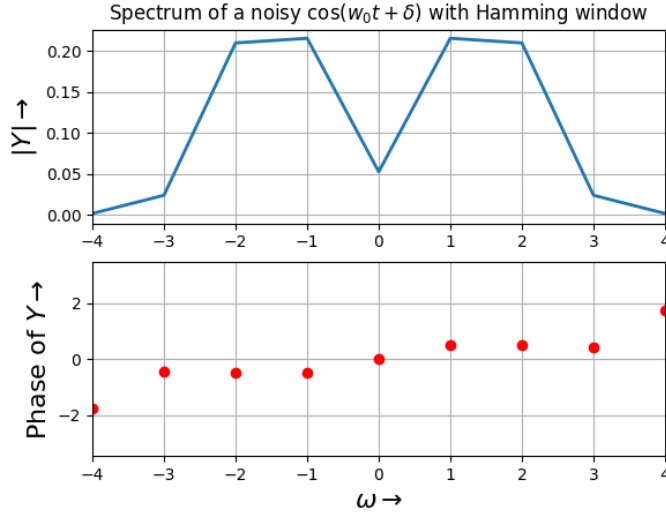


Figure 9: Fourier transform of noise + $\cos(1.5t + 0.5)$

3.4 Question 5

In this question we analyze a chirp signal which is an FM signal where frequency is directly proportional to time. A chirp signal we shall consider is given by

$$f(t) = \cos(16t(1.5 + \frac{t}{2\pi})) \quad (2)$$

The FFT of the chirp is given by: We note that the frequency response is spread between 5-50 rad/s. A large section of this range appears due to Gibbs phenomenon. On windowing, only frequencies between 16 and 32 rad/s remain.

```
t = linspace(-pi,pi,1025); t = t[: -1]
dt = t[1] - t[0]; fmax = 1/dt
n = arange(1024)
wnd = fftshift(0.54 + 0.46 * cos(2 * pi * n / 1024))
y = cos(16 * t * (1.5 + t / (2 * pi))) * wnd
y[0] = 0
y = fftshift(y)
Y = fftshift(fft(y)) / 1024.0
w = linspace(-pi * fmax, pi * fmax, 1025); w = w[: -1]
plot_spectrum(6, w, Y, 100, r" Spectrum of chirped function", r"$|Y|$ \rightarrow")
```

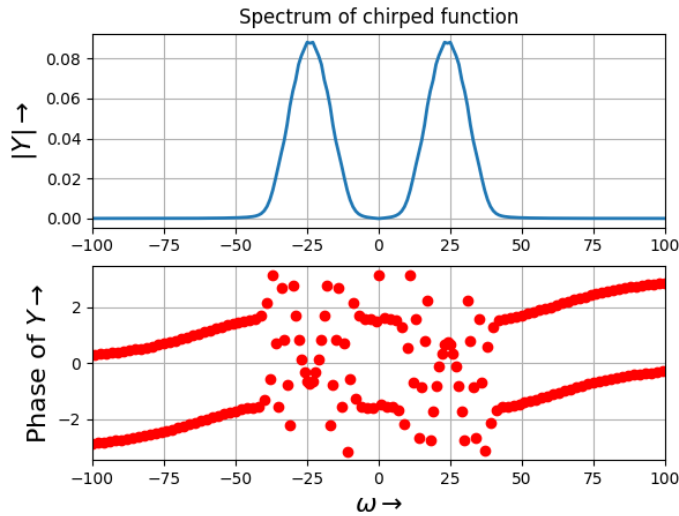


Figure 10: Chirp function fourier transform, windowed

3.5 Question 6

For the same chirped signal, we break the 1024 vector into pieces that are 64 samples wide. Extract the DFT of each and store as a column in a 2D array. Then plot the array as a surface plot to show how the frequency of the signal varies with time.

```
t_array = split(t,16)
Y_mag = zeros((16,64))
Y_phase = zeros((16,64))

for i in range(len(t_array)):
    n = arange(64)
    wnd = fftshift(0.54+0.46*cos(2*pi*n/64))
    y = cos(16*t_array[i]*(1.5 + t_array[i]/(2*pi)))*wnd
    y[0]=0
    y = fftshift(y)
    Y = fftshift(fft(y))/64.0
    Y_mag[i] = abs(Y)
    Y_phase[i] = angle(Y)

t = t[:,64]
w = linspace(-fmax*pi,fmax*pi,64+1); w = w[:-1]
t,w = meshgrid(t,w)

fig1 = figure(7)
```

```

ax = fig1.add_subplot(111, projection='3d')
surf=ax.plot_surface(w,t,Y_mag.T,cmap='viridis',linewidth=0, antialiased=
fig1.colorbar(surf, shrink=0.5, aspect=5)
ax.set_title('surface plot');
ylabel(r"$\omega \rightarrow$")
xlabel(r"$t \rightarrow$")

```

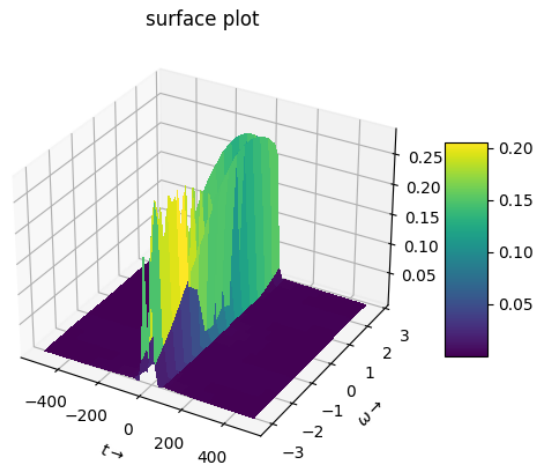


Figure 11: Chopped Chirp function