

HW4_4&5

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Cogs 118C HW4

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1 4

```
[1]: import numpy as np
import matplotlib.pyplot as plt
```

1.1 a.

Answer

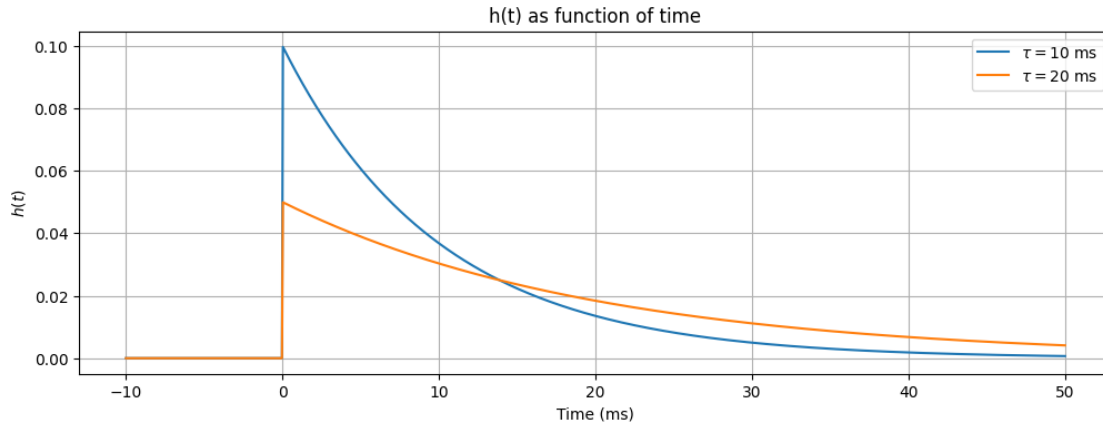
```
[17]: def h(t, tau):
        return (1/tau) * (t >= 0) * np.exp(-t / tau)

t = np.linspace(-10, 50, 1000)
tau_values = [10, 20]

plt.figure(figsize = (12, 4))

for tau in tau_values:
    plt.plot(t, h(t, tau), label = f'$\\tau = {tau}$ ms')

plt.title('h(t) as function of time')
plt.xlabel('Time (ms)')
plt.ylabel('$h(t)$')
plt.legend()
plt.grid(True)
plt.show()
```



1.2 b

Answer > increases = slower impulse response

decreases = faster impulse response

1.3 c

Answer > $H(\omega) = \int_{-\infty}^{\infty} h(t)e^{-i\omega t} dt$

$$= \int_0^{\infty} \frac{1}{\tau} e^{-\frac{t}{\tau}} e^{-i\omega t} dt = \frac{1}{1 + i\omega\tau}$$

1.4 d

Answer > $g(\omega) = \left| \frac{1}{1 + i\omega\tau} \right| = \frac{1}{\sqrt{1 + (\omega\tau)^2}} = \frac{1}{\sqrt{1 + (0 \cdot \tau)^2}} = \frac{1}{\sqrt{1 + 0}} = \frac{1}{1} = 1$

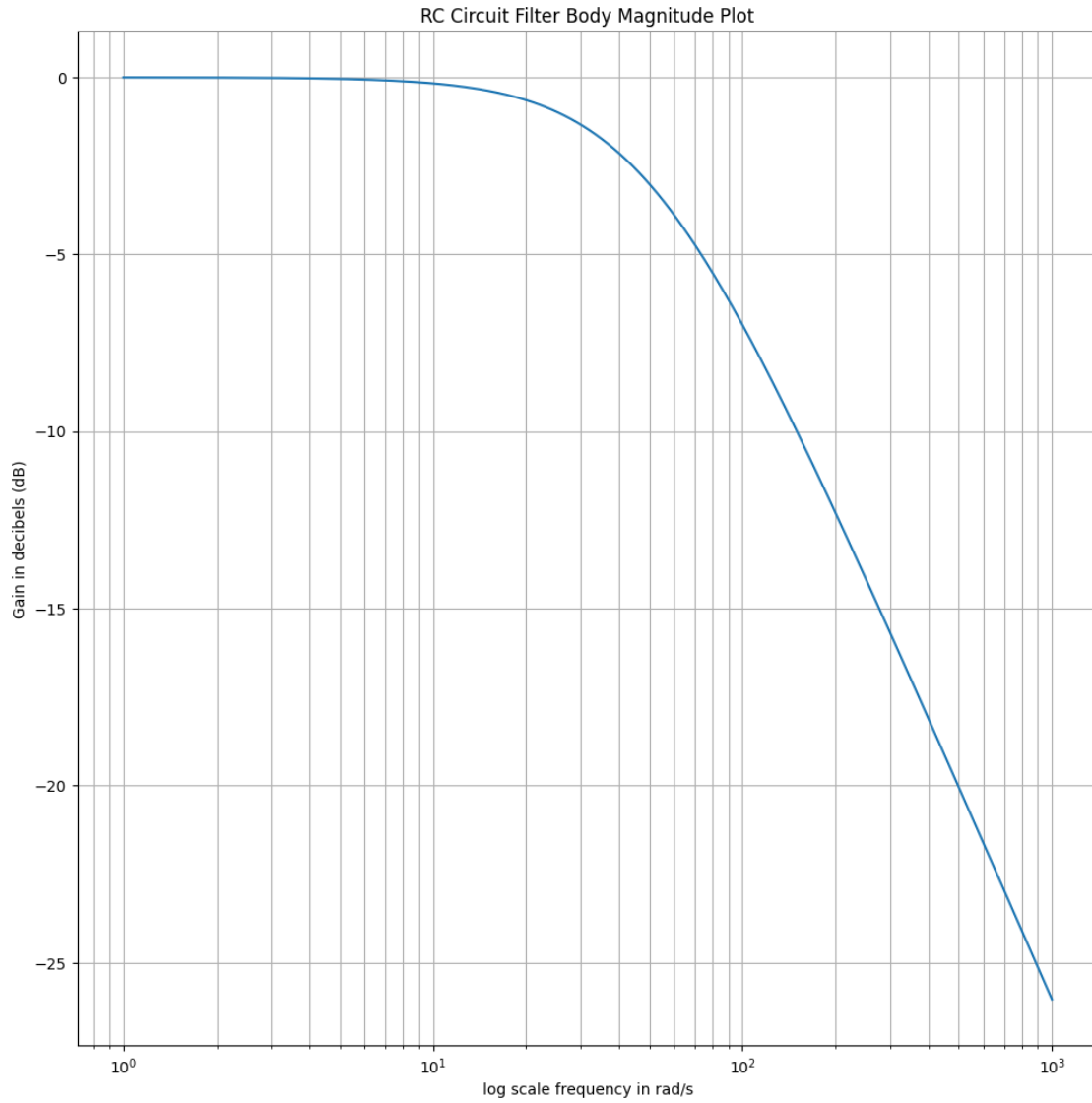
1.5 e

Answer

```
[21]: tau_20ms = 20e-3
omega_range = np.logspace(0, 3, 300)
omega_g_function_gain = 1 / np.sqrt(1 + (omega_range * tau_20ms)**2)

omega_g_function_db = 20 * np.log10(omega_g_function_gain)

plt.figure(figsize = (12, 12))
plt.semilogx(omega_range, omega_g_function_db)
plt.title('RC Circuit Filter Body Magnitude Plot')
plt.xlabel('log scale frequency in rad/s')
plt.ylabel('Gain in decibels (dB)')
plt.grid(True, which = "both", ls = "-")
plt.show()
```



1.6 f

Answer $> 20 \log_{10} \frac{1}{\sqrt{1 + (2\pi \cdot 60 \cdot 0.02)^2}} = -17.622952655178295 \text{ db}$

- $20 \text{ ms} \cdot \frac{1}{1000} = 0.02 \text{ s}$
- Angular function: $w = \frac{2\pi}{T} \rightarrow f_{\text{frequency}} = \frac{1}{T} \rightarrow w = \Theta = 2\pi f$
- – T = period

1.7 g

$$\text{Im}[H(\omega)] = -\frac{1}{1+(w\tau)^2}$$

$$\text{Re}[H(\omega)] = \frac{1}{1+(w\tau)^2}$$

$$\tan\phi(\omega) = \frac{\text{Im}[H(\omega)]}{\text{Re}[H(\omega)]} = \frac{-\frac{1}{1+(w\tau)^2}}{\frac{1}{1+(w\tau)^2}} = -1 \rightarrow -w\tau = (2\pi \cdot 60)(0.02) = 7.53982236$$

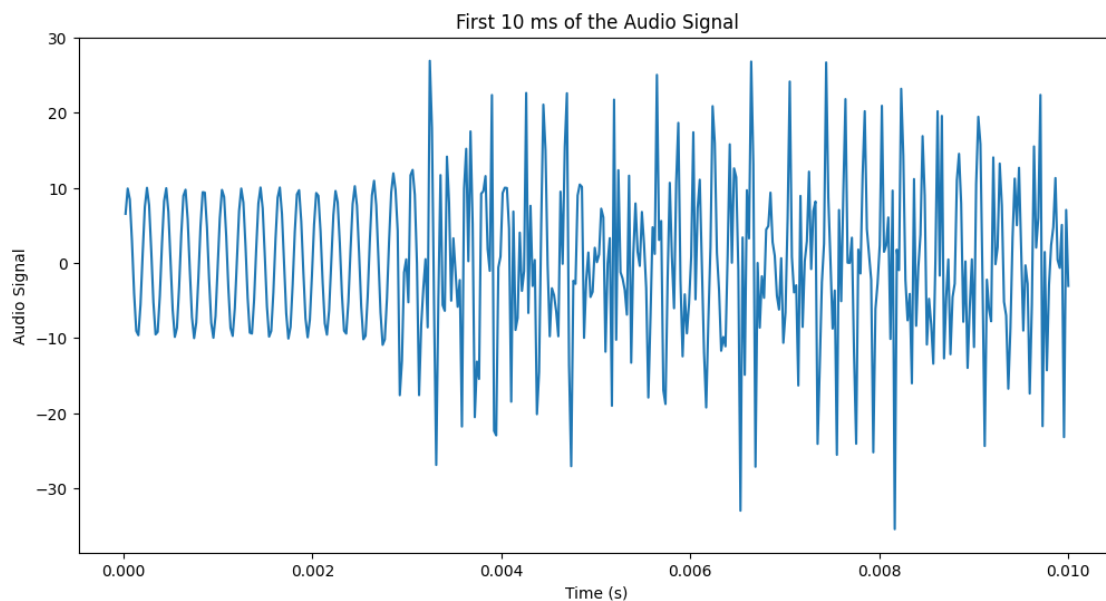
Answer > 7.53982236

2 5

```
[2]: # Load the audio file
data = np.loadtxt('hw4_audio.csv',delimiter=',')
t = data[:,0]
x = data[:,1]
Fs = 44100 # Sampling rate in Hz

[3]: # (a) Save a .wav audio file. You should be able to download
# this file and play it to your speakers. How does it sound?
from scipy.io import wavfile
wavfile.write('hw4_audio.wav',Fs,x)

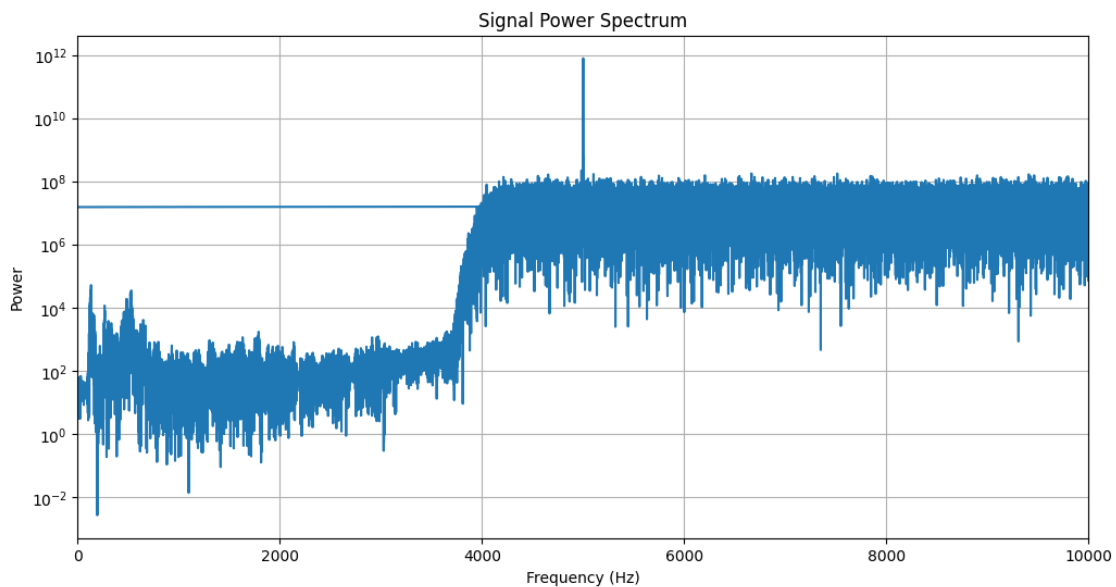
[4]: # (b) Make a plot showing the first 10 ms of the signal.
# Do you notice a strong sinusoidal oscillation?
plt.figure(figsize = (12, 6))
first_10ms = 0.01
number_of_samples_in_10ms = Fs * first_10ms
time_range = int(number_of_samples_in_10ms)
plt.plot(t[0:time_range], x[0:time_range])
plt.title('First 10 ms of the Audio Signal')
plt.xlabel('Time (s)')
plt.ylabel('Audio Signal')
plt.show()
```



```
[5]: # (c) Plot the power spectrum of the signal. Use
# log scale for the y-axis. Use plt.xlim() to zoom
# in on the frequency range from 0 to 10 kHz. Finally,
# make sure to label the axes.
#
# Hint: use np.fft.fft, np.fft.fftfreq
```

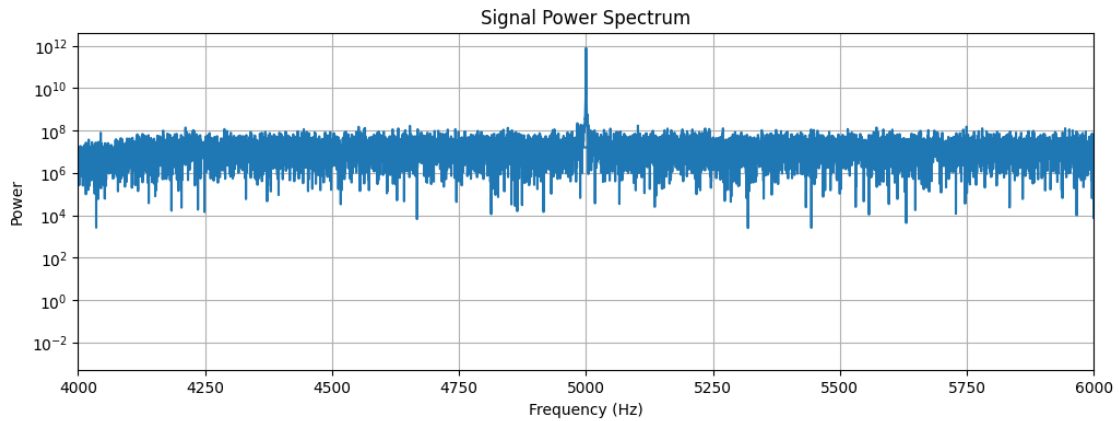
```
X = np.fft.fft(x)
f = np.fft.fftfreq(len(x), 1 / Fs)

plt.figure(figsize = (12, 6))
plt.plot(f, np.abs(X)**2)
plt.yscale('log')
plt.xlim(0, 10000)
plt.title('Signal Power Spectrum')
plt.xlabel('Frequency (Hz)')
plt.ylabel('Power')
plt.grid(True)
plt.show()
```

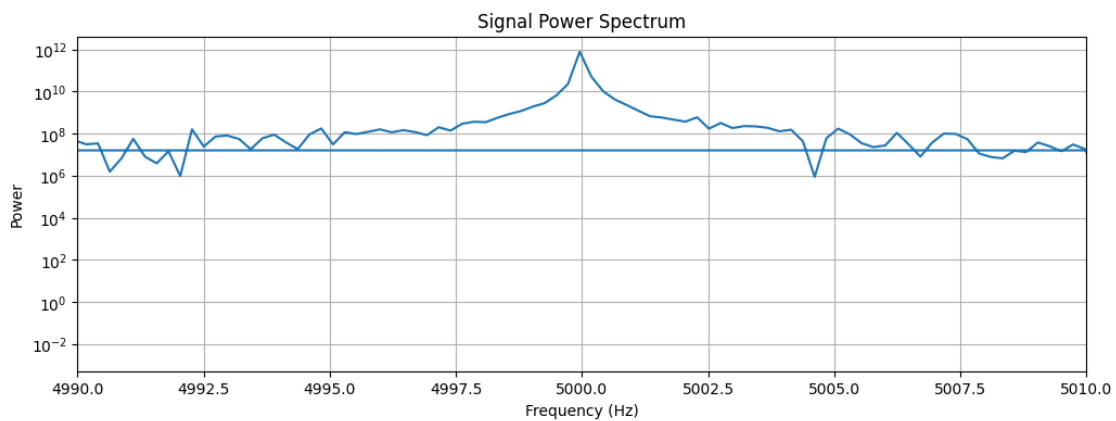


```
[6]: plt.figure(figsize = (12, 4))
plt.plot(f, np.abs(X)**2)
plt.yscale('log')
plt.xlim(4000, 6000)
plt.title('Signal Power Spectrum')
```

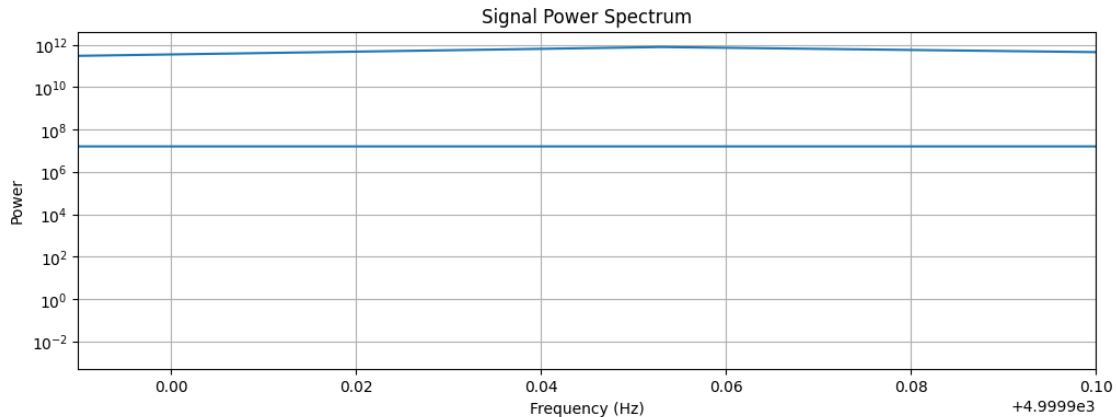
```
plt.xlabel('Frequency (Hz)')
plt.ylabel('Power')
plt.grid(True)
plt.show()
```



```
[7]: plt.figure(figsize = (12, 4))
plt.plot(f, np.abs(X)**2)
plt.yscale('log')
plt.xlim(4990, 5010)
plt.title('Signal Power Spectrum')
plt.xlabel('Frequency (Hz)')
plt.ylabel('Power')
plt.grid(True)
plt.show()
```



```
[8]: plt.figure(figsize = (12, 4))
plt.plot(f, np.abs(X)**2)
plt.yscale('log')
plt.xlim(4999.89, 5000)
plt.title('Signal Power Spectrum')
plt.xlabel('Frequency (Hz)')
plt.ylabel('Power')
plt.grid(True)
plt.show()
```



- 3 (d) What is the exact frequency of the largest peak
- 4 in the power spectrum? Hint: You may need to zoom in even
- 5 further on the plot you made in (c) to see exactly where
- 6 the peak is.

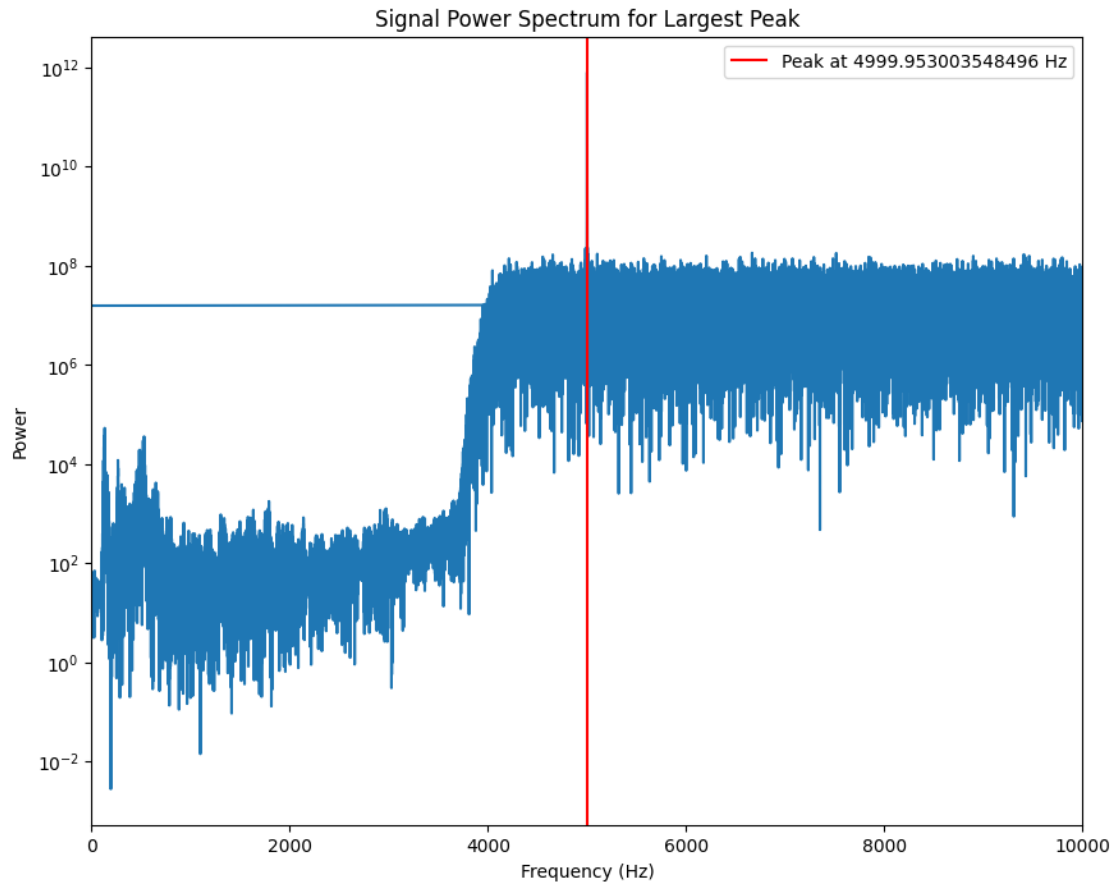
Answer > 4999.953003548496 Hz

```
[9]: largest_peak_index = np.argmax(np.abs(X[:len(f)//2])**2)
largest_peak_frequency = f[largest_peak_index]
print("The exact frequency of the largest peak is: ", largest_peak_frequency, "
      ↪ Hz")
```

The exact frequency of the largest peak is: 4999.953003548496 Hz

```
[10]: plt.figure(figsize = (10, 8))
plt.plot(f, np.abs(X)**2)
plt.axvline(x = largest_peak_frequency, color = 'r', linestyle = '-', label = "
      ↪ Peak at {largest_peak_frequency} Hz")
plt.yscale('log')
```

```
plt.xlim(0, 10000)
plt.title('Signal Power Spectrum for Largest Peak')
plt.xlabel('Frequency (Hz)')
plt.ylabel('Power')
plt.legend()
plt.show()
```



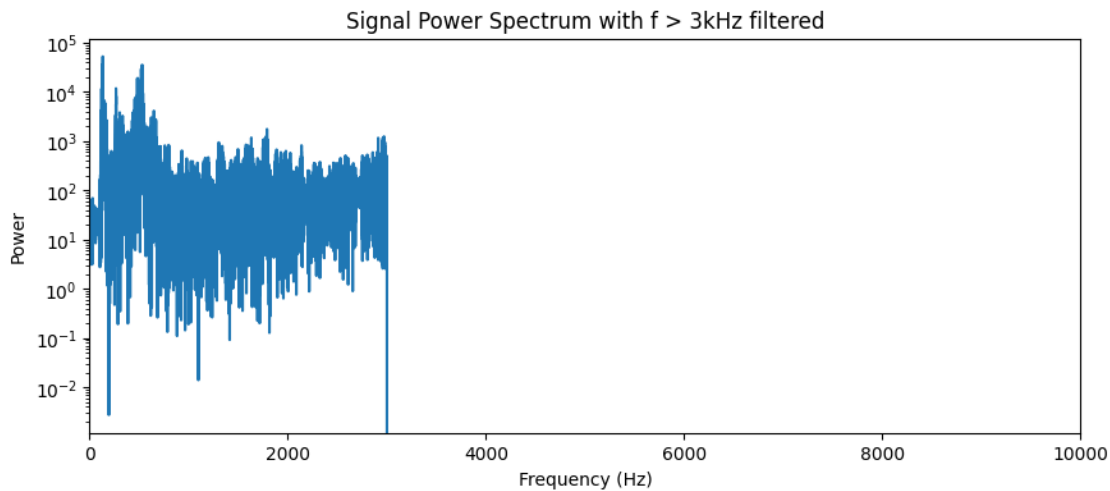
```
[11]: # (e) Ideal filter: Make a filter, H, that will remove frequencies
# f > 3 kHz. Plot the product of H with X, the Fourier transform of x

# To get you started, here is how to create H:
H = np.abs(f) < 3000
filtering_X = X * H

plt.figure(figsize = (10, 4))
plt.plot(f, np.abs(filtering_X)**2)
plt.yscale('log')
plt.xlim(0, 10000)
plt.title('Signal Power Spectrum with f > 3kHz filtered')
```



```
plt.xlabel('Frequency (Hz)')
plt.ylabel('Power')
plt.show()
```



Plot e shows a limited range of the original audio and filters frequencies above 3 kHz

```
[12]: # (f) Use the convolution theorem to calculate the filtered signal.
# To do this, note that the filtered signal is the inverse Fourier transform
# (ifft) of the product of the Fourier transforms of x and h.
# Save a .wav audio file containing the filtered signal. Download and listen to
# it.
# Describe what you hear.

# Hint: use wavfile.write

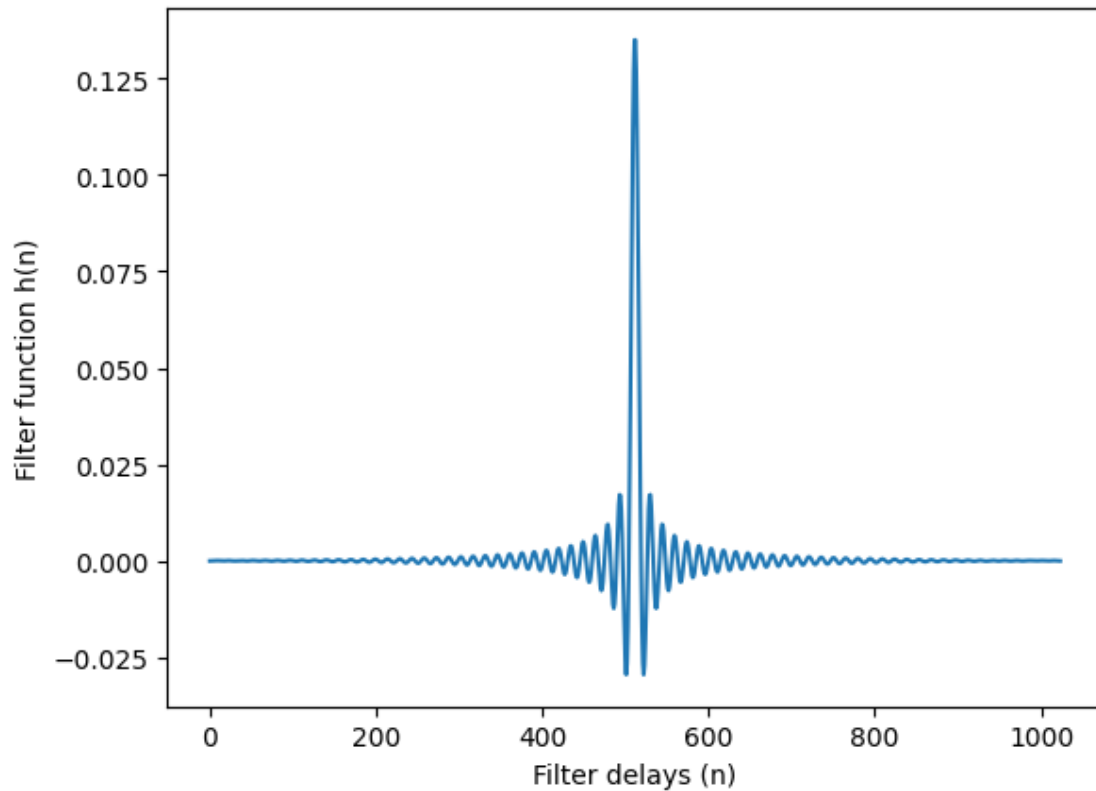
filtered_x = np.real(np.fft.ifft(filtering_X))
wavfile.write('hw4_audio_filtered.wav', Fs, filtered_x)
```

```
[13]: # (g) Apply the high-pass filter we designed to remove the high frequencies
# from the data. You may use np.convolve, with mode='same' so that the output
# has the same size as the input. Plot the power spectrum of the filtered
# signal,
# again showing the range 0 to 10 kHz. Comment on the similarity and difference
# of this spectrum from the original spectrum (part c).
```

```
[14]: # Design a better high-pass filter using a finite impulse response (FIR)
# convolution. We will create a filter with order n=1024
from scipy import signal
n = 1024
h = signal.firwin(n,cutoff=3000,fs=Fs) # This function designs a FIR filter
```

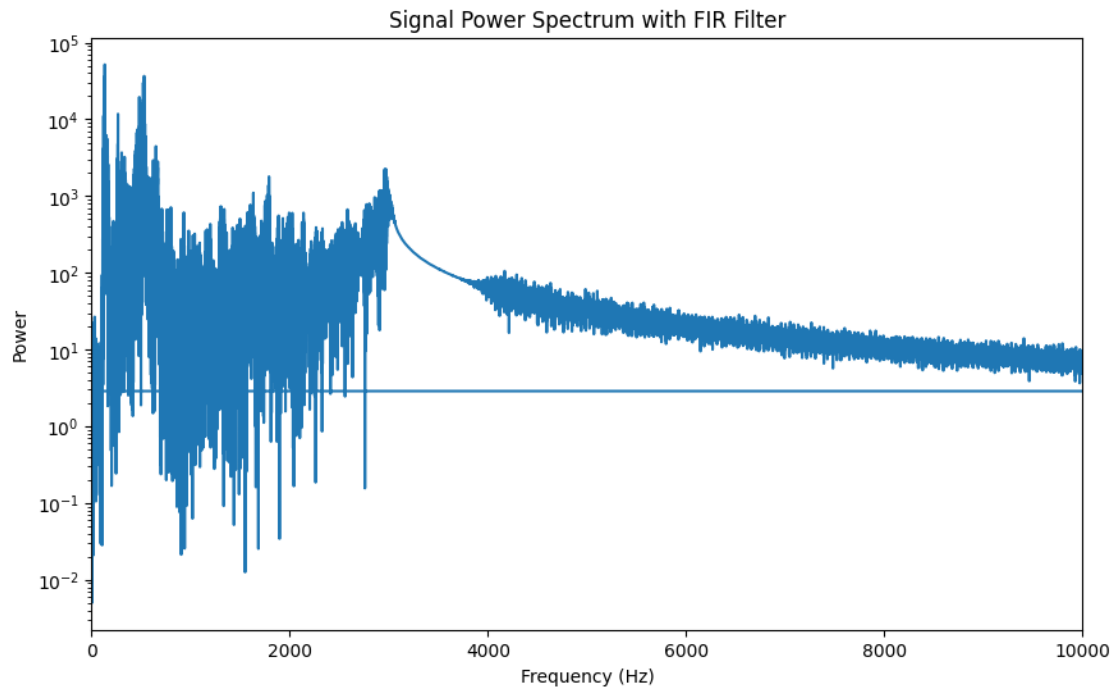
```
plt.plot(h)
plt.xlabel('Filter delays (n)')
plt.ylabel('Filter function h(n)')
```

[14]: Text(0, 0.5, 'Filter function h(n)')



```
[15]: filtered_x_fir = np.convolve(x, h, mode = 'same')
filtered_X_fir = np.fft.fft(filtered_x_fir)

plt.figure(figsize=(10, 6))
plt.plot(f, np.abs(filtered_X_fir)**2)
plt.yscale('log')
plt.xlim(0, 10000)
plt.title('Signal Power Spectrum with FIR Filter')
plt.xlabel('Frequency (Hz)')
plt.ylabel('Power')
plt.show()
```



```
[16]: # (h) Save the filtered signal as a .wav file and listen
      # to it on your speakers. What is the "secret" message
      # (actually a quote from the eminent UCSD neuro-philosopher
      # Patricia Churchland)?

      wavfile.write('hw4_audio_lp_filter.wav', Fs, filtered_x_fir))
```