

# Design and Application of an Elliptic Filter (w/ Python)

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**TCET 3102-E316 (Analog and Digital Com) Lab 1**

**Spring 2019, Section: E316, Code: 37251**

**Instructor: Song Tang**

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**Date: 4/08/2019**

**TCET 3102-E316 (Analog and Digital Communications) Lab 3**

**Spring 2019, Section: E316, Code: 37251**

**Instructor: Song Tang**

## Objective

- Design a low pass elliptic filter to receive desired output (filtered signal), where the input will be a square wave signal.

## Equipment

- Computer Software

## Theory

- The subscriber loop is the most critical component of a telephone network. They're created through the use of two copper wires.
- We can create an equivalent circuit for the subscriber loop through the use of a low pass filter. The bandwidth of this filter can be calculated w/ the following formula:

$$\omega_c = \frac{1}{RC} = 0.707$$

- 0.707 is the maximum magnitude and the cutoff frequency
- For telephone networks, the bandwidth of the subscriber loop is enough to support signal transmissions of the range **0 - 4 kHz**. Also, to further improve quality of transmissions, designers use loading coils to flatten attenuation-frequency characters of the loop. This was how telephone networks used to function.
- A problem arises, however, with the arrival of the digital era. Since digital signals are pulse waveforms they have a spectrum much larger than the range mentioned above. The objective of our experiment is to determine the response of digital signals through the subscriber loops.

## Modules (Packages)

```
In [1]: 1 # These are the packages I'll need to solve this problem
        2 import math as ma
        3 import numpy as np
        4 from matplotlib import pyplot as plt
        5 from scipy.fftpack import fft, fftfreq
        6 from scipy.signal import ellip, freqz, lfilter, freqs
```

## Variables for time domain plot

The following function can also be used to generate a fourier series

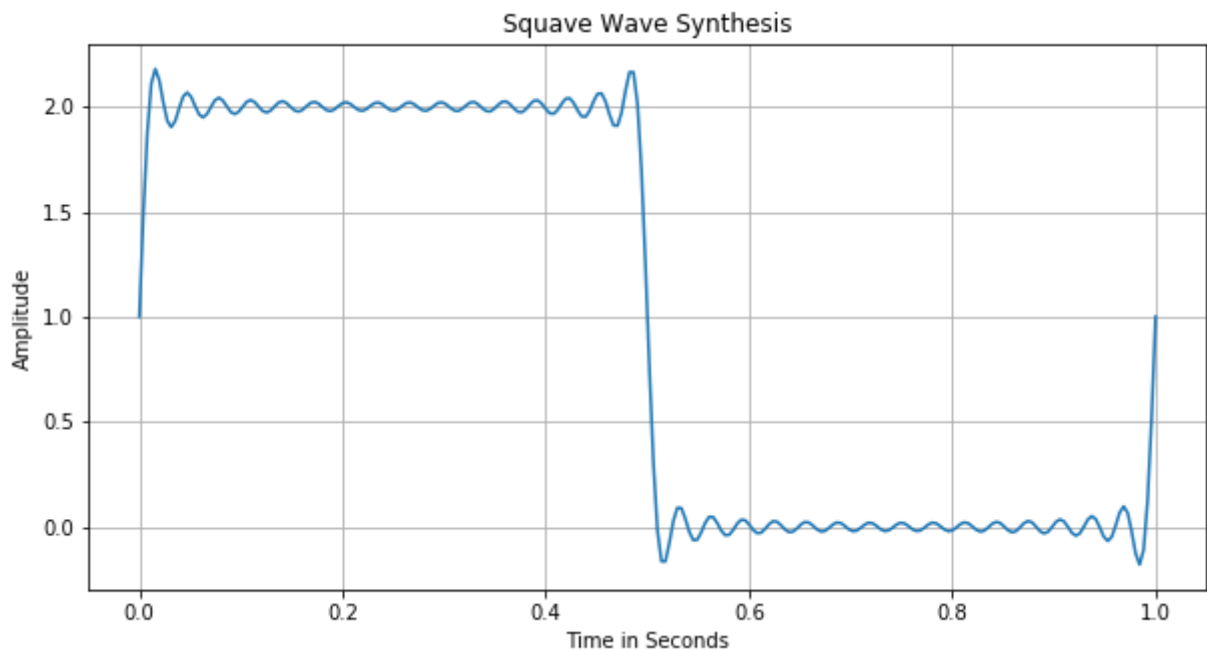
```
def ffs_(omega, t, A):
    s = np.sin(omega*t)
    for i in range(3, 30, 2):
        s = s + 1/i * np.sin(i*omega*t)
    s = A/2 + (2*A)/np.pi * s
    return s
```

```
In [2]: 1 # These variables are used to create the fourier series
        2 # Start and Stop indicates my domain for my sampling frequency [sampling
        3 # 100 points.
        4 # n1, n2 are the amount of harmonics I want.
        5 start, stop, n, samplingfreq = 0, 100, 15, 100
        6
        7 # 't' is for time, and is used to create my 100 Hz time vector
        8 t = np.linspace(start, stop, 256) / samplingfreq
        9
        10 # 'A' represents the amplitude 2 volts
        11 A = 2
        12
        13 # 'fundamental' is the DC component of the Fourier Series
        14 fundamental = A/2
        15
        16 # 'signalfreq' is the Signal Frequency (f_0)
        17 signalfreq = 1
        18
        19 # 'omega' is the Angular Velocity (w_0)
        20 omega = 2 * np.pi * signalfreq
        21
        22 # Lambda function
        23 template = lambda p: ((2*A)/(np.pi*(2*p+1))) * np.sin((2*p+1) * omega *
        24
        25 # harmonics1, harmonics2 are AC component of the Fourier series
        26 harmonics = sum([template(p) for p in range(n+1)])
        27
        28 # ffs1, ffs2 are the fourier series
        29 ffs = lambda n: (fundamental + harmonics)
```

## RUN 1: Synthesis Square Wave

Step 1: This is using the square wave from Lab 1

```
In [3]: 1 # Creates figure 1 and its subplot
2 fig1, ax1 = plt.figure(figsize= (10,5)), plt.subplot()
3 ax1.plot(t, ffs(n))
4 ax1.set(xlabel= 'Time in Seconds', ylabel= 'Amplitude',
5         title= 'Squave Wave Synthesis');
6 ax1.grid(True)
```



The following code gives very similar results as the one shown above

```
# Creates figure 2 and its subplot
# This here uses a different fourier series function.
# I wanted to see if results later down the lab would be different
fig2, ax2 = plt.figure(figsize= (10,5)), plt.subplot()
ax2.plot(t, ffs_(omega, t, A))
ax2.set(xlabel= 'Frequency (Hertz)', ylabel= 'Amplitude',
       title= 'Squave Wave Synthesis');
ax2.grid(True)
```

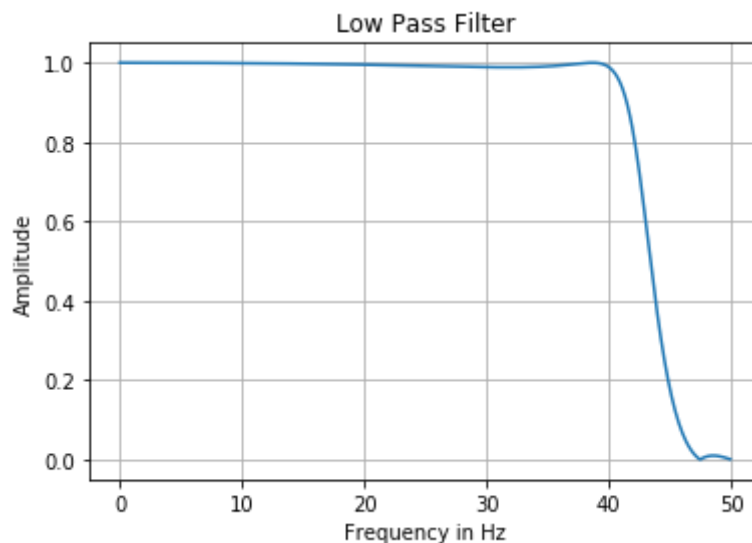
## RUN 2: Elliptic (Cauer) Filter Design

### Step 1: Plot the Elliptic (Cauer) Filter

```

In [4]: 1 # Elliptical order, 'm'
2 m = 3
3
4 # Decibels of the ripple in passband, 'rP'
5 rP = 0.1
6
7 # Decibels of the ripple in stopband, 'rS'
8 rS = 40
9
10 # The Cutoff Filter Frequency, 'omega_N', normalized to Nyquist Frequency
11 # 'omega_N = 1' corresponds to half the sampling frequency
12 # When the frequency equals to 'omega_N' the filter's magnitude response
13 # equal to the ripple in passband, 'rP' (decibels)
14
15 omega_N = 0.8
16
17 # in-built scipy function for the Elliptic (Cauer) Filter
18 b, a = ellip(m, rP, rS, omega_N)
19
20 # displays the magnitude and phase of the filter, normalized to Nyquist
21 omega_, H = freqz(b, a)
22 xval = (omega_ * samplingfreq)/(2 * np.pi)
23 yval = abs(H)
24 plt.plot(xval, yval)
25 plt.xlabel('Frequency in Hz')
26 plt.ylabel('Amplitude')
27 plt.title('Low Pass Filter')
28 plt.grid()

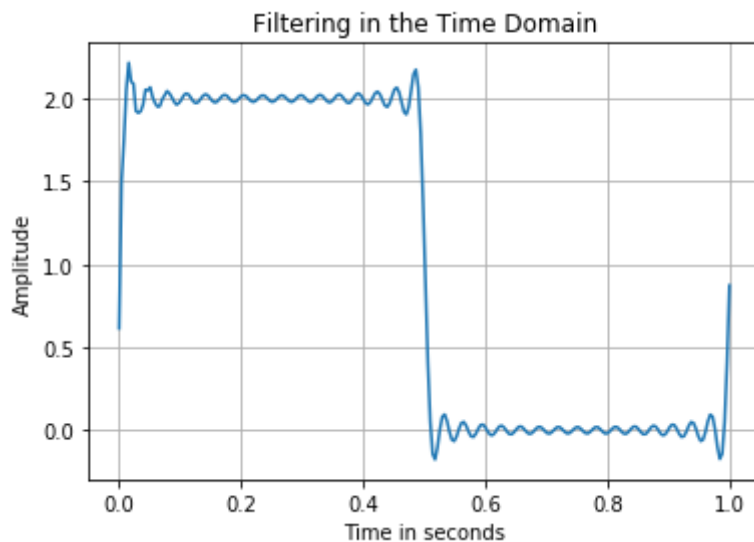
```



## RUN 3: Filter Input Signals

### Step 1: Observe the filtering effect in the time domain

```
In [5]: 1 sf_ = lfilter(b, a, ffs(n))
2 plt.plot(t, sf_)
3 plt.xlabel('Time in seconds')
4 plt.ylabel('Amplitude')
5 plt.title('Filtering in the Time Domain')
6 plt.grid()
```



**Also gives similar results as the one above**

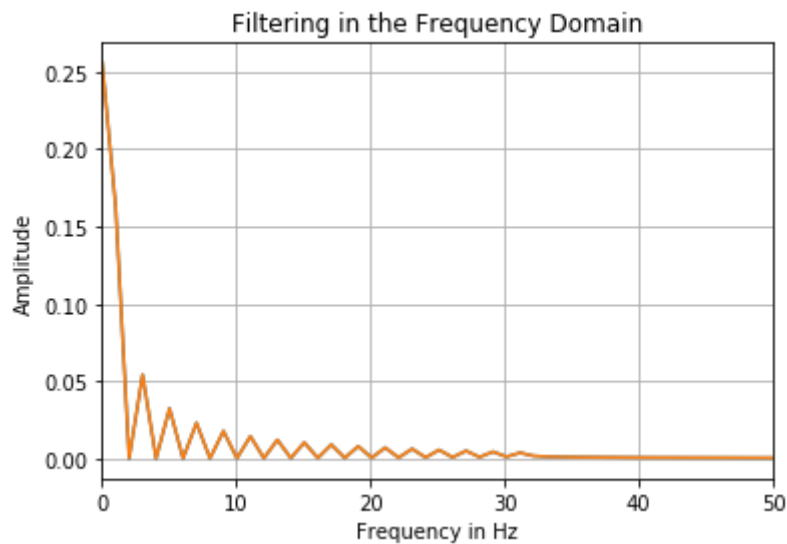
```
sf = lfilter(b, a, ffs_(omega, t, A))
plt.plot(t, sf)
```

**Step 2: Observe the filtering effect in the frequency domain**

**Also gives similar results as the one below**

```
S = fft(ffs_(omega, t, A))[:256]/1000
SF = fft(sf)[:256]/1000
f = np.linspace(0,256, 256)
var1, var2 = abs(S), abs(SF)
plt.plot(f, var1)
plt.plot(f, var2)
plt.xlim(0,50)
plt.grid()
```

```
In [6]: 1 S = fft(ffs(n))[:256]/1000
2 SF = fft(sf_)[:256]/1000
3 f = np.linspace(0,256, 256)
4 var1, var2 = abs(S), abs(SF)
5 plt.plot(f, var1)
6 plt.plot(f, var2)
7 plt.xlim(0,50)
8 plt.xlabel('Frequency in Hz');
9 plt.ylabel('Amplitude')
10 plt.title('Filtering in the Frequency Domain')
11 plt.grid()
```



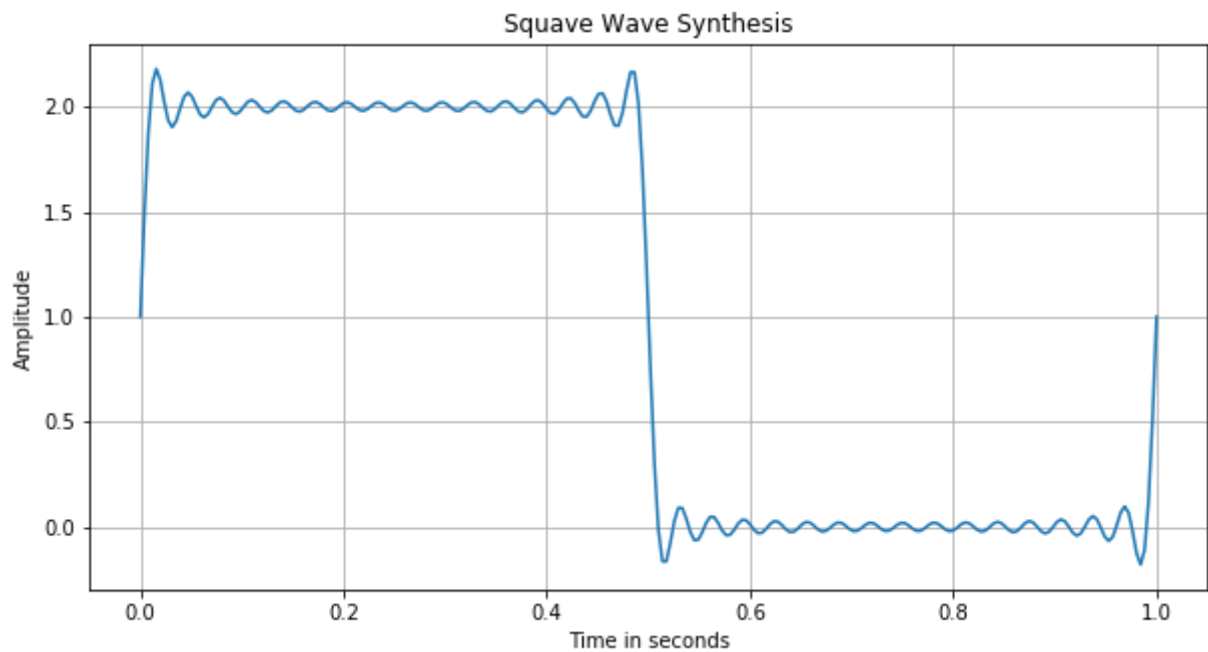
## Lab Questions/Requirements

**Question 1: Follow Run 1 and synthesis a square wave with  $A = 2\text{ V}$  and  $T = 1\text{ ms}$ \*\***

- Check RUN 1 for the answer, as it is already shown w/ the same initial conditions



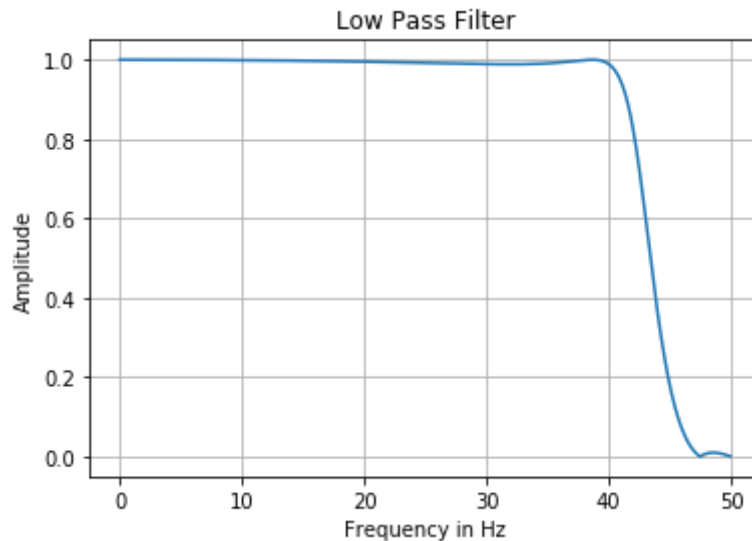
```
In [7]: 1 # Creates figure 1 and its subplot
2 fig1, ax1 = plt.figure(figsize= (10,5)), plt.subplot()
3 ax1.plot(t, ffs(n))
4 ax1.set(xlabel= 'Time in seconds', ylabel= 'Amplitude',
5         title= 'Squave Wave Synthesis');
6 ax1.grid(True)
```



**Question 2: Follow RUN 2 and design a low pass filter w/ a bandwidth of 4 kHz. Depict the Magnitude-Frequency and Phase-Frequency characteristics**

- Check RUN 2 for the answer, as it is already shown w/ the same initial conditions

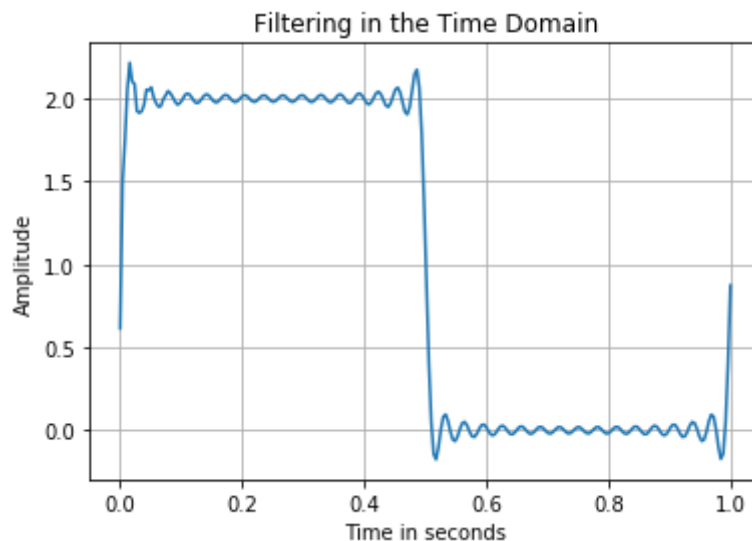
```
In [8]: 1 plt.plot(xval, yval)
2 plt.xlabel('Frequency in Hz')
3 plt.ylabel('Amplitude')
4 plt.title('Low Pass Filter')
5 plt.grid()
```



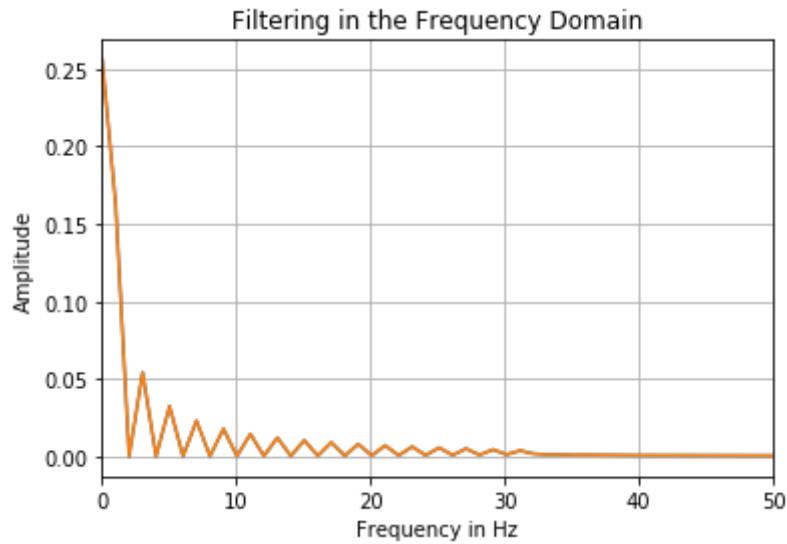
**Question 3: Now pass the square wave through the lowpass filter and depict the time and frequency domain response**

- Check RUN 3 for the answer, as it is already shown w/ the same initial conditions

```
In [9]: 1 sf_ = lfilter(b, a, ffs(n))
2 plt.plot(t, sf_)
3 plt.xlabel('Time in seconds')
4 plt.ylabel('Amplitude')
5 plt.title('Filtering in the Time Domain')
6 plt.grid()
```



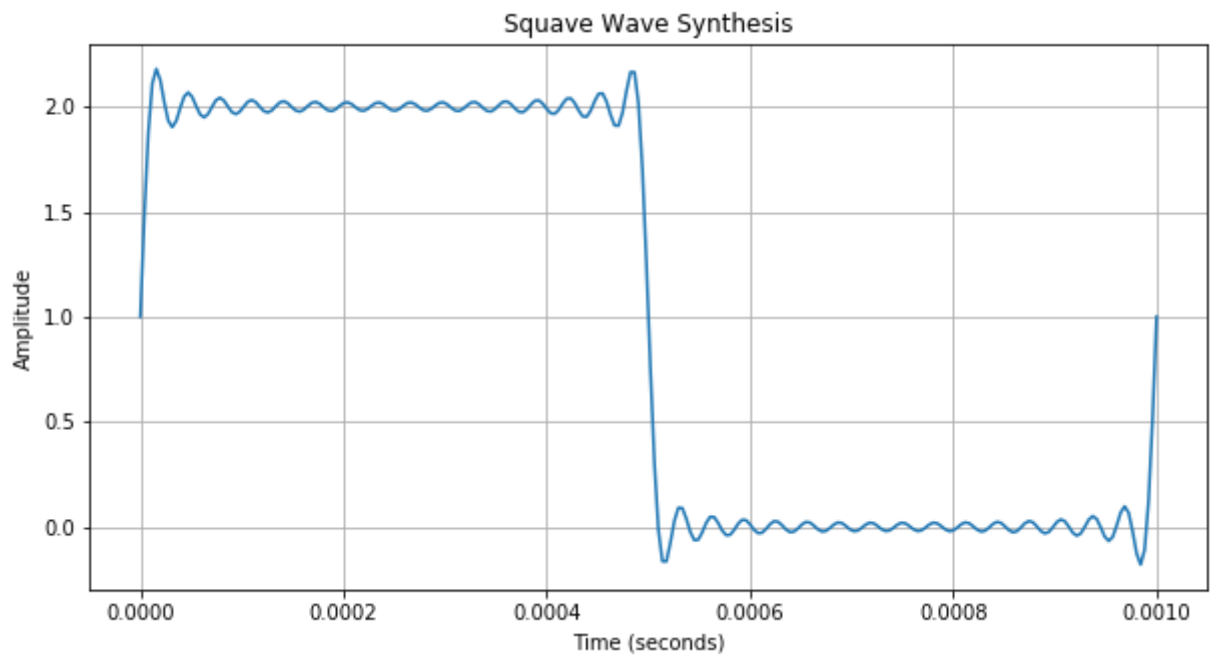
```
In [10]: 1 plt.plot(f, var1)
2         plt.plot(f, var2)
3         plt.xlim(0,50)
4         plt.xlabel('Frequency in Hz');
5         plt.ylabel('Amplitude')
6         plt.title('Filtering in the Frequency Domain')
7         plt.grid()
```



**Question 4: Modify the filter design to pass only the first harmonic of the input square wave. Show results in the time and frequency domain**

```
In [11]: 1 # These variables are used to create the fourier series
2 # Start and Stop indicates my domain for my sampling frequency [samplin
3 # 100 points.
4 # n1, n2 are the amount of harmonics I want.
5 start, stop, n, samplingfreq = 0, 100, 15, 100000
6
7 # 't' is for time, and is used to create my 100 Hz time vector
8 t = np.linspace(start, stop, 256) / samplingfreq
9
10 # 'A' represents the amplitude 2 volts
11 A = 2
12
13 # 'fundamental' is the DC component of the Fourier Series
14 fundamental = A/2
15
16 # 'signalfreq' is the Signal Frequency (f_0)
17 signalfreq = 1000
18
19 # 'omega' is the Angular Velocity (w_0)
20 omega = 2 * np.pi * signalfreq
21
22 # Lambda function
23 template = lambda p: ((2*A)/(np.pi*(2*p+1))) * np.sin((2*p+1) * omega *
24
25 # harmonics1, harmonics2 are AC component of the Fourier series
26 harmonics = sum([template(p) for p in range(n+1)])
27
28 # ffs1, ffs2 are the fourier series
29 ffs = lambda n: (fundamental + harmonics)
```

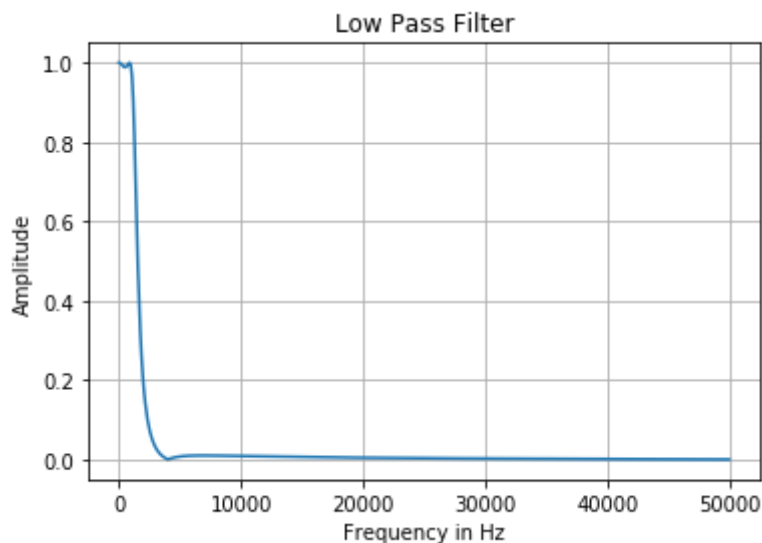
```
In [12]: 1 # Creates figure 1 and its subplot
2 fig1, ax1 = plt.figure(figsize= (10,5)), plt.subplot()
3 ax1.plot(t, ffs(n))
4 ax1.set(xlabel= 'Time (seconds)', ylabel= 'Amplitude',
5         title= 'Squave Wave Synthesis');
6 ax1.grid(True)
```



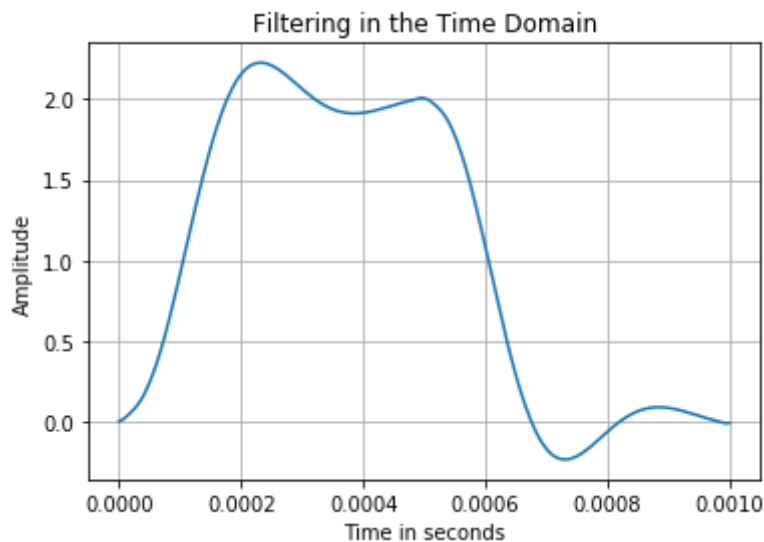
```

In [13]: 1 # Elliptical order, 'm'
2 m = 3
3
4 # Decibels of the ripple in passband, 'rP'
5 rP = 0.1
6
7 # Decibels of the ripple in stopband, 'rS'
8 rS = 40
9
10 # The Cutoff Filter Frequency, 'omega_N', normalized to Nyquist Frequency
11 # 'omega_N = 1' corresponds to half the sampling frequency
12 # When the frequency equals to 'omega_N' the filter's magnitude response
13 # equal to the ripple in passband, 'rP' (decibels)
14
15 omega_N = 0.02
16
17 # in-built scipy function for the Elliptic (Cauer) Filter
18 b_, a_ = ellip(m, rP, rS, omega_N)
19
20 # displays the magnitude and phase of the filter, normalized to Nyquist
21 omega_, H = freqz(b_, a_)
22 xval = (omega_ * samplingfreq)/(2 * np.pi)
23 yval = abs(H)
24 plt.plot(xval, yval)
25 plt.xlabel('Frequency in Hz')
26 plt.ylabel('Amplitude')
27 plt.title('Low Pass Filter')
28 plt.grid()

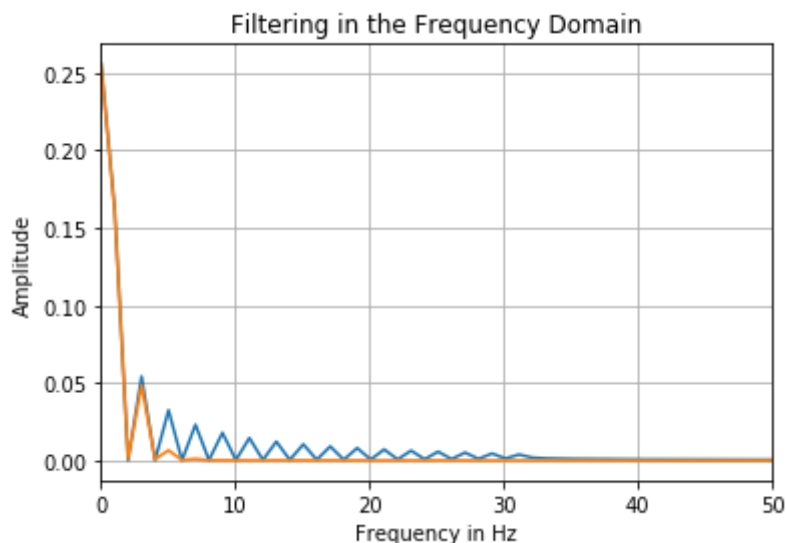
```



```
In [14]: 1 sf_ = lfilter(b_, a_, ffs(n))
2         plt.plot(t, sf_)
3         plt.xlabel('Time in seconds')
4         plt.ylabel('Amplitude')
5         plt.title('Filtering in the Time Domain')
6         plt.grid()
```



```
In [15]: 1 S = fft(ffs(n))[:256]/1000
2         SF = fft(sf_)[:256]/1000
3         f = np.linspace(0,256, 256)
4         var1, var2 = abs(S), abs(SF)
5         plt.plot(f, var1)
6         plt.plot(f, var2)
7         plt.xlim(0,50)
8         plt.xlabel('Frequency in Hz');
9         plt.ylabel('Amplitude')
10        plt.title('Filtering in the Frequency Domain')
11        plt.grid()
```



**Question 5: Repeat the previous step to pass only the three first harmonics of the input square wave. Show results in the time and frequency domain.**

```

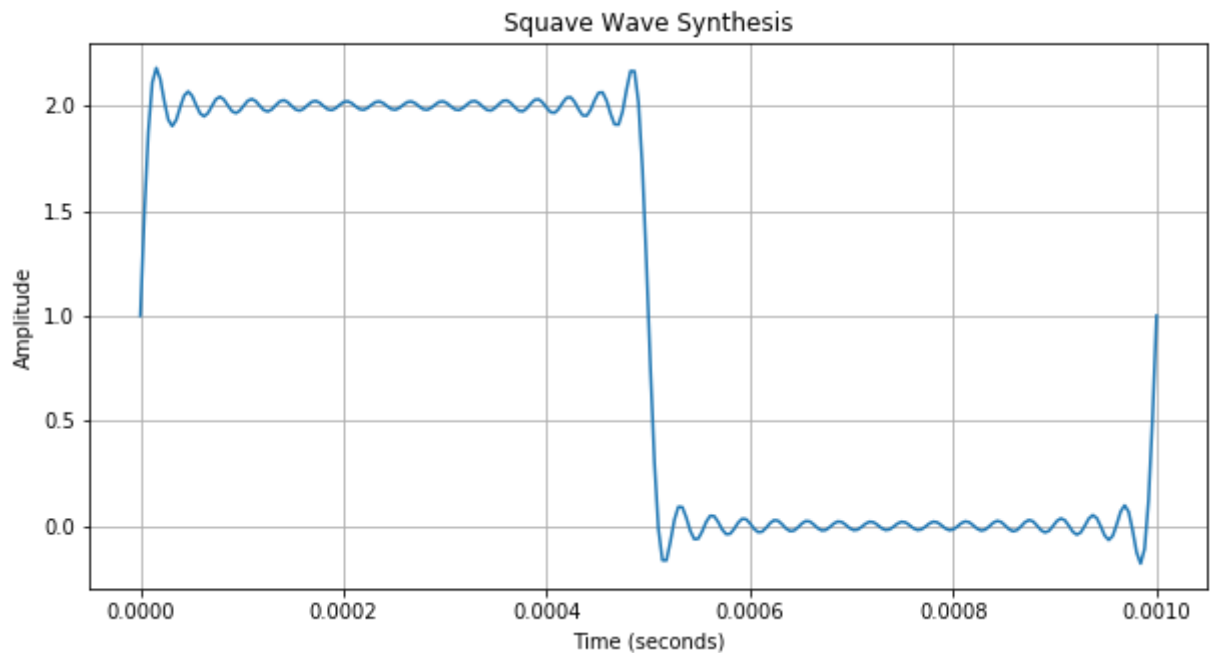
In [16]: 1 # These variables are used to create the fourier series
2 # Start and Stop indicates my domain for my sampling frequency [samplin
3 # 100 points.
4 # n1, n2 are the amount of harmonics I want.
5 start, stop, n, samplingfreq = 0, 100, 15, 100000
6
7 # 't' is for time, and is used to create my 100 Hz time vector
8 t = np.linspace(start, stop, 256) / samplingfreq
9 # 'A' represents the amplitude 2 volts
10 A = 2
11 # 'fundamental' is the DC component of the Fourier Series
12 fundamental = A/2
13 # 'signalfreq' is the Signal Frequency (f_0)
14 signalfreq = 1000
15 # 'omega' is the Angular Velocity (w_0)
16 omega = 2 * np.pi * signalfreq
17 # Lambda function
18 template = lambda p: ((2*A)/(np.pi*(2*p+1))) * np.sin((2*p+1) * omega *
19 # harmonics1, harmonics2 are AC component of the Fourier series
20 harmonics = sum([template(p) for p in range(n+1)])
21 # ffs1, ffs2 are the fourier series
22 ffs = lambda n: (fundamental + harmonics)

```

```

In [17]: 1 # Creates figure 1 and its subplot
2 fig1, ax1 = plt.figure(figsize= (10,5)), plt.subplot()
3 ax1.plot(t, ffs(n))
4 ax1.set(xlabel= 'Time (seconds)', ylabel= 'Amplitude',
5         title= 'Squave Wave Synthesis');
6 ax1.grid(True)

```

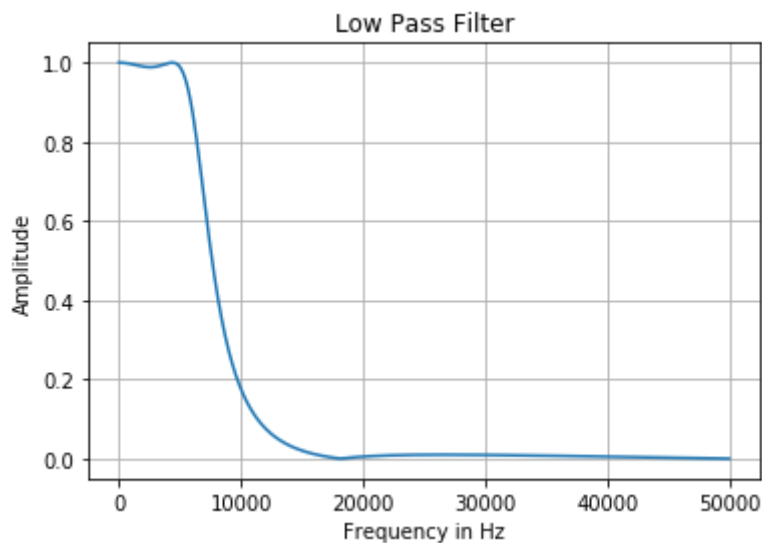




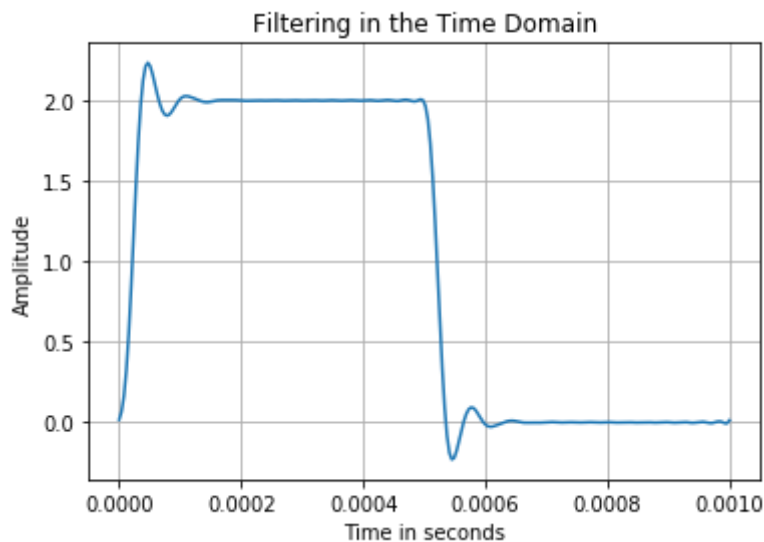
```

In [18]: 1 # Elliptical order, 'm'
2 m = 3
3 # Decibels of the ripple in passband, 'rP'
4 rP = 0.1
5 # Decibels of the ripple in stopband, 'rS'
6 rS = 40
7 # The Cutoff Filter Frequency, 'omega_N', normalized to Nyquist Frequency
8 # 'omega_N = 1' corresponds to half the sampling frequency
9 # When the frequency equals to 'omega_N' the filter's magnitude response
10 # equal to the ripple in passband, 'rP' (decibels)
11 omega_N = 0.1
12 # in-built scipy function for the Elliptic (Cauer) Filter
13 b, a = ellip(m, rP, rS, omega_N)
14 # displays the magnitude and phase of the filter, normalized to Nyquist
15 omega_, H = freqz(b, a)
16 xval = (omega_ * samplingfreq)/(2 * np.pi)
17 yval = abs(H)
18 plt.plot(xval, yval)
19 plt.xlabel('Frequency in Hz')
20 plt.ylabel('Amplitude')
21 plt.title('Low Pass Filter')
22 plt.grid()

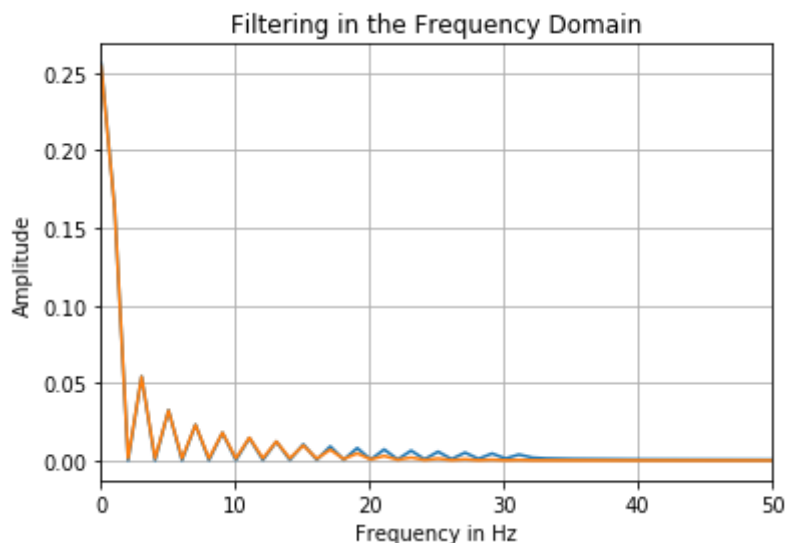
```



```
In [19]: 1 sf_ = lfilter(b, a, ffs(n))
2         plt.plot(t, sf_)
3         plt.xlabel('Time in seconds')
4         plt.ylabel('Amplitude')
5         plt.title('Filtering in the Time Domain')
6         plt.grid()
```



```
In [20]: 1 S = fft(ffs(n))[:256]/1000
2         SF = fft(sf_)[:256]/1000
3         f = np.linspace(0,256, 256)
4         var1, var2 = abs(S), abs(SF)
5         plt.plot(f, var1)
6         plt.plot(f, var2)
7         plt.xlim(0,50)
8         plt.xlabel('Frequency in Hz');
9         plt.ylabel('Amplitude')
10        plt.title('Filtering in the Frequency Domain')
11        plt.grid()
```



**Question 6: If you design an R-C Circuit ( $R = 10\text{ k}\Omega$  and  $C = 1\text{ pF}$ ) with a resistor in series and a capacitor in parallel with output terminals, what would be the filter bandwidth and why**

**it would work as a low pass filter.**

- This would work as an frequencies higher than 16 MHz would get attenuated

```
In [21]: 1 cutoff = 1/(2*np.pi*R*C)
          2 print('Cutoff Frequency (Bandwidth): ',cutoff(10*ma.pow(10,3), 1*ma.p

Cutoff Frequency (Bandwidth):  15.915494309189533 MHz
```

## Analysis/Conclusion

I was somewhat successful with this lab. We determined that an increase/decrease in the sampling frequency and fundamental frequency by a factor of 1000 will not change properties of the filter. The scaling will be the only thing magnified. Changing the filter cutoff frequency to pass the first harmonic causes us to have a sine wave time domain like response. Basically the filtered signal looks almost like a sine wave. When we change the filter cutoff frequency to pass three harmonics, we have more ripples, according to our graphs. The first harmonic tends to roll off faster than when passing the third harmonics.