Exploring Aliasing

Setting up an aliased wave

In [169]:

import numpy as np
import cmath

import matplotlib.pyplot as plt

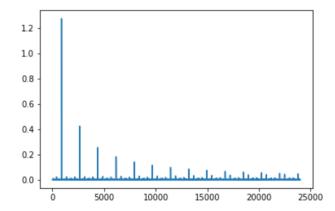
```
In [170]:
          class waveGen:
                   def init (self, freg, BufferSize, waveType, vol):
                           self.sampleRate = 48000
                           self.bufferSize = BufferSize
                           self.buffer = np.empty(BufferSize, dtype=np.com
          plex)
                           self.frequency = freq
                           self.omega = 0
                           self.updateOmega()
                           self.last = complex(1,0)
                           self.type = waveType
                           self.volume = vol
                   def sineGen(self):
                           #start = timer() #reduce function!!! functools
                           for i in range(0, self.bufferSize):
                                   self.buffer[i] = self.last
                                   self.last=self.last*self.omega
                           return self.volume * np.imag(self.buffer).astyp
          e(np.float32)
                   def squareGen(self):
                           #start = timer()
                           for i in range(0, self.bufferSize):
                                   self.buffer[i] = np.sign(self.last)
                                   self.last=self.last*self.omega
                           return np.real(self.volume * self.buffer).astyp
          e(np.float32)
                   def updateOmega(self):
                           self.omega = cmath.exp(1j*(2*cmath.pi * self.fr
          equency / self.sampleRate))
                   def updateFreq(self, val):
                           self.frequency = val
                           self.updateOmega()
                   def updateFPS(self, val):
                           self.sampleRate = val
                           self.updateOmega()
```

```
In [171]: sampleRate = 48000
    length = 48000
    freq = 880
    wave = waveGen(freq, length, 'square', 1)

In [172]: square = wave.nextFrame()
    freqSpec = np.fft.fft(square)
    #linespaceing
    T = 1.0 / sampleRate

#x angle of plot
    xf = np.linspace(0.0, 1.0/(2.0*T), length//2)

#ploting the graph
    plt.plot(xf, 2.0/length * np.abs(freqSpec[:length//2]))
    plt.show()
```



in a pure square wave we would expect to see peaks on every other harmonic

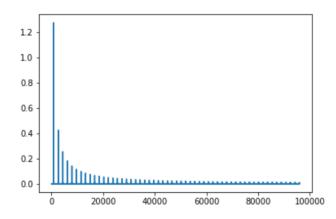
whareas we can see here that the harmonics have looped back on themselfs, creating a background noise of around .05 decebals across the spectrum.

This distortion is allready baked into the wave, so we can try the same again by increasing the FPS, therby increasing the nyquist window

The Over Sample wave

```
In [173]:
           #an 8 times over sampled wave
           overSampling = 4
           wave.updateFPS(sampleRate*overSampling)
           wave.updateBuffSize(length*overSampling)
           #generate data
           square2 = wave.nextFrame()
           freqSpec2 = np.fft.fft(square2)
           #linespaceing
           T = 1.0 / (sampleRate*overSampling)
           #x angle of plot
           xf = np.linspace(0.0, 1.0/(2.0*T), length * overSampling//2)
           #ploting the graph
           plt.plot(xf, 2.0/(length*overSampling) * np.abs(freqSpec2[:leng
           th * overSampling//2]))
           plt.show()
```

/usr/lib/python3.6/site-packages/ipykernel_launcher.py:27: ComplexWarning: Casting complex values to real discards the imaginary part



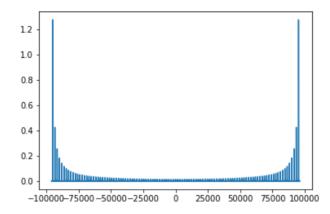
it can be seen here that the distortion is much less apparent.

the nyquist value here should be at 96,000 HZ

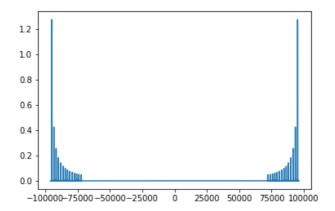
It is notable that the distortion here is higher in the higher range

the full data of the fft looks like this, and may provide a better idea of how to downSample

In [174]: #linespaceing T = 1.0 / (sampleRate*overSampling) #x angle of plot xf = np.linspace(-length*2, 1.0/(2.0*T), length*overSampling) #ploting the graph plt.plot(xf, 2.0/(length*overSampling) * np.abs(freqSpec2)) plt.show()

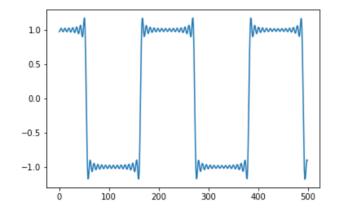


to filter here at 24000 HZ we want to nullify all values bettween -24000 and 24000



we can return this to the time domain and then downSample

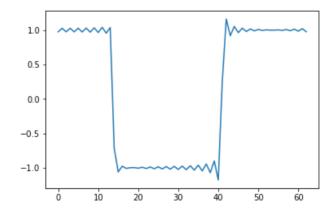
```
In [176]: fltSqr = np.fft.ifft(np.real(freqSpec3))
    plt.plot(np.real(fltSqr[:500]))
    plt.show()
```



```
In [177]: downSampled = np.empty(sampleRate)

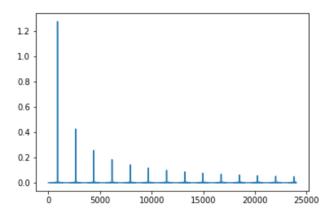
for i in range(len(fltSqr)):
    if i % overSampling == 0:
        downSampled[i//overSampling] = np.real(fltSqr[i])

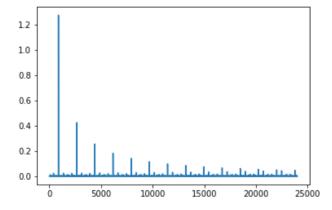
plt.plot(downSampled[:63])
    plt.show()
```



lets veiw this in the frequecy domain, and compare with our original graph

```
In [178]:
          DSFreqSpec = np.fft.fft(downSampled)
           #linespaceing
           T = 1.0 / (sampleRate)
           #x angle of plot
           xf = np.linspace(0.0, 1.0/(2.0*T), sampleRate // 2)
           #ploting the down sampled graph
           plt.plot(xf, 2.0/length * np.abs(DSFreqSpec[:length // 2]))
           plt.show()
           #linespaceing
           T = 1.0 / length
           #x angle of plot
           xf = np.linspace(0.0, 1.0/(2.0*T), length//2)
           #ploting the original graph
           plt.plot(xf, 2.0/length * np.abs(freqSpec[:length//2]))
           plt.show()
```

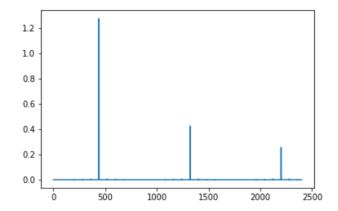




there is clearly still some Distortion, but there is at least much less

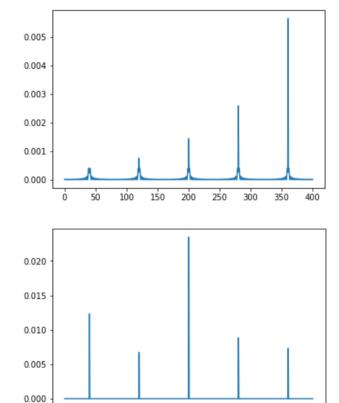
heres a zoom on a distored area to show more clearly

```
In [179]: zoom = 10
    #linespaceing
T = 1.0 / length * zoom
#x angle of plot
xf = np.linspace(0.0, 1.0/(2.0*T), length//zoom)
#ploting the original graph
plt.plot(xf, 2.0/length * np.abs(DSFreqSpec[:length//zoom]))
plt.show()
```



and a little closer... focusing on the distortion, with the original signal distortion plotted below

```
In [185]: zoom = 60
    #linespaceing
T = 1.0 / length * zoom
    #x angle of plot
    xf = np.linspace(0.0, 1.0/(2.0*T), length//zoom)
    #ploting the original graph
    plt.plot(xf, 2.0/length * np.abs(DSFreqSpec[:length//zoom]))
    plt.show()
    plt.plot(xf, 2.0/length * np.abs(freqSpec[:length//zoom]))
    plt.show()
```



50

Ó

100

150

200

250

300

it can be seen here that by using this method a background distortion is prevelent across the spectrum of ~0.005 decibal at a 4 times overSample rate

350

400

thereby reducing the distortion levels by aproximatly 4 times

This method is however computatioaly complex and far from perfect

a priority should be in avoiding the FFT in filter design, and finding methods that can avoid high levels of over sampling