

Method 1

Noise filtering and then deblurring the given transmitted signal $y[n]$

In [24]:

```
# importing required libraries for basic mathematics and manipulation

import pandas as pd
import matplotlib.pyplot as plt
import numpy as np
import math
```

In [25]:

```
#reads the csv and stores it in a dataframe

df = pd.read_csv('data.csv')
```

In [26]:

```
#displays first five rows
df.head()
```

Out[26]:

	x[n]	y[n]
0	35.4312	33.3735
1	35.1511	34.3744
2	34.8284	35.7514
3	34.4656	35.5869
4	34.0656	36.0826

In [27]:

```
# removing headers
df = df.rename(columns={col: "" for col in df})
df.head()
```

Out[27]:

0	35.4312	33.3735
1	35.1511	34.3744
2	34.8284	35.7514
3	34.4656	35.5869
4	34.0656	36.0826

In [28]:

```
# converting data into 1D lists
x = df.iloc[:,0]
y = df.iloc[:,1]
y.head()
```

Out[28]:

```
0    33.3735
1    34.3744
2    35.7514
3    35.5869
4    36.0826
Name: , dtype: float64
```

In [29]:

```
# converts to numpy array which supports complex number manipulation
x = np.array(x)
y = np.array(y)
```

Representation of initial signals

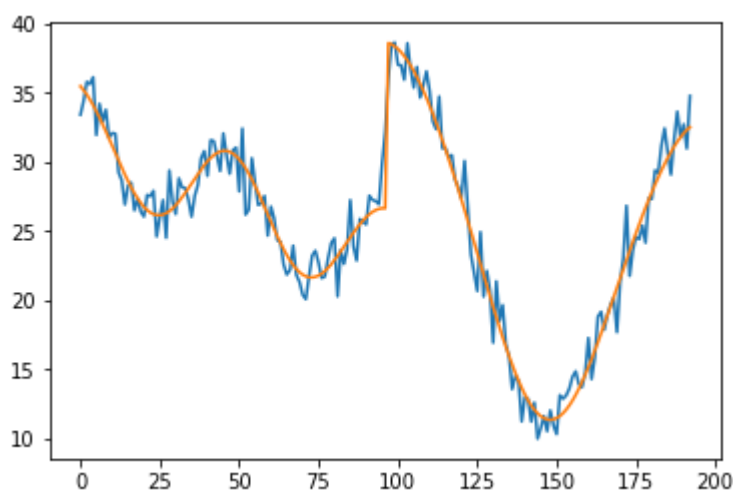
The orange plot represents original signal $x[n]$ and blue plot represents transmitted signal $y[n]$

In [30]:

```
plt.plot(range(0,193),y)
plt.plot(range(0,193),x)
```

Out[30]:

[<matplotlib.lines.Line2D at 0x28b07adb8b0>]



Step 1 : Removing noise by using moving averages

In [31]:

```

window_size = 5          # indicates that we are taking average of 5 consecutive samples

i = 0
y_denoised = []

while i < len(y) - window_size + 1:          # while loop used untill the array ends
                                              # one unit everytime the loop runs

    this_window = y[i : i + window_size]

    window_average = sum(this_window) / window_size
    y_denoised.append(window_average)
    i += 1

y_denoised.insert(-2, (y[-2]+y[-3]+y[-4])/3)          # we fall short of 4 values when window
y_denoised.insert(-1, (y[-1]+y[-2]+y[-3])/3)          # manually by using the same method but
                                              # minimum deviation at the edges

y_denoised.insert(0, (y[0]+y[1]+y[2])/3)
y_denoised.insert(1, (y[1]+y[2]+y[3])/3)

```

Plot of the denoised signal against original signal

In [32]:

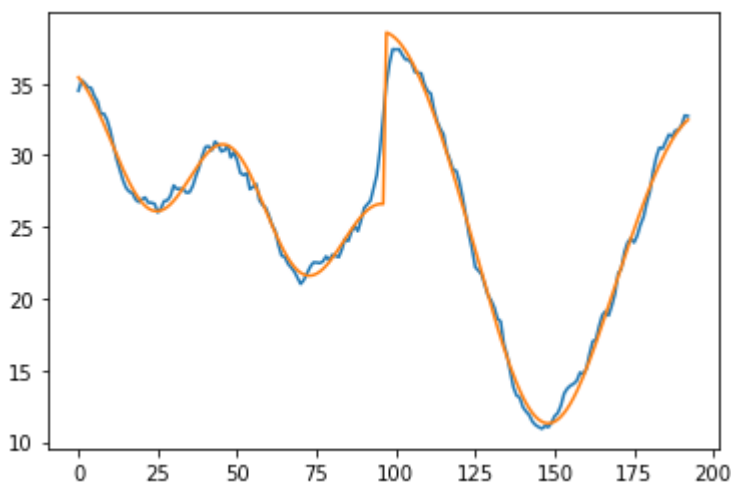
```

plt.plot(range(0,193),y_denoised)
plt.plot(range(0,193),x)

```

Out[32]:

[<matplotlib.lines.Line2D at 0x28b08baef70>]



Deblur using DTFT

In [33]:

```

# dtft algorithm for the denoised version of y[n]

N = len(y_denoised)
w = np.linspace(0,2*math.pi,N)          # taking 193 samples of w between 0 and 2*pi

X = np.zeros(N,dtype = 'complex_')      # array to store dtft

for k in range(0,N-1):                  # outer loop to change the value of k
    X[k] = 0

    for n in range(0,N-1):              # inner loop to change the value of n
        X[k] = X[k] + y_denoised[n]* math.e**(-1j *w[k] *n)          # formula for dtf

```

In [34]:

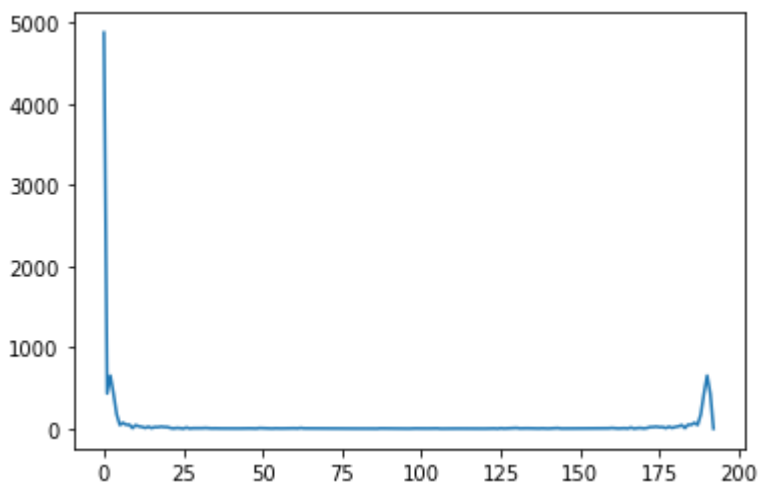
```

# plot for dtft
plt.plot(range(0,193),abs(X))

```

Out[34]:

[<matplotlib.lines.Line2D at 0x28b08c106d0>]



In [35]:

```

# dtft for h[n]

h = np.array([1/16, 1/4, 3/8, 1/4, 1/16])          # given

N = 193

w=[]
for i in range (193):                             # taking 193 samples of w between 0 and
    w.append(2*math.pi*i/193)

H = np.zeros(N,dtype = 'complex_')

for k in range(0,N-1):                             # same dtft algorithm as above
    H[k-2] = 0                                     # replace k by 2-k since the middle value

    for n in range(0,5):

        H[k-2] = H[k-2] + h[n]* math.e**(-1j *w[k-2] *n)

```

Plot for DTFT of h[n]

In [36]:

```

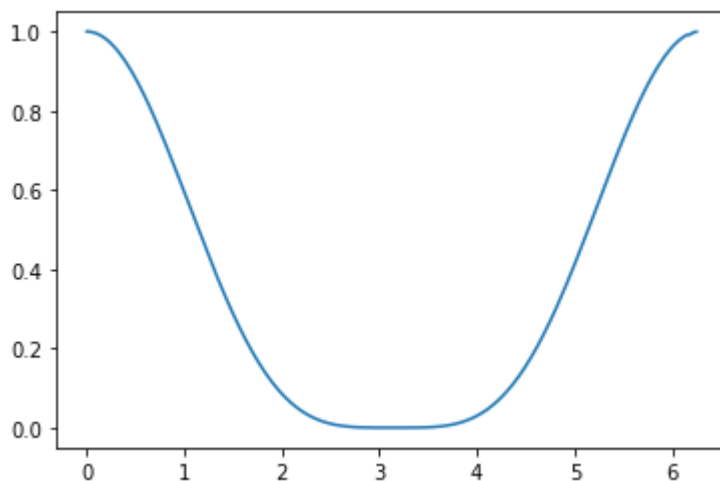
H[190] = (H[191] + H[189]) /2

plt.plot(range(0,193),abs(H))

```

Out[36]:

[<matplotlib.lines.Line2D at 0x28b08c74250>]



In [37]:

```
# In the above plot we see that  $H(e^{j\omega})$  approaches zero at certain values and thus  $X_{final}(e^{j\omega})$ 
# To avoid this scenario we clip the graph at an appropriate value which is determined by threshold
# value to all the values below the threshold
```

```
H_mod = list(abs(H))
```

```
for n, i in enumerate(H_mod):
    if i < 0.8:
        H_mod[n] = 0.8
```

In [38]:

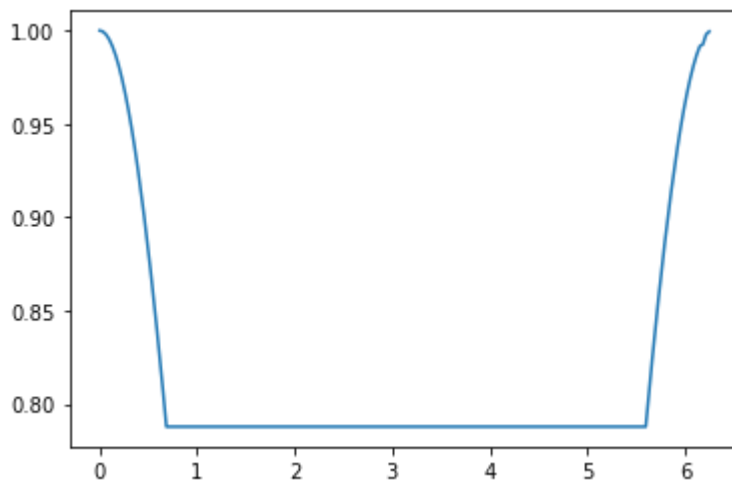
```
for i in range(22,172):           # assigns the given to all the numbers in this range
    H[i] = 0.15920910143656403-0.7716376423250288j
```

In [39]:

```
plt.plot(range(0,193),abs(H))
```

Out[39]:

```
[<matplotlib.lines.Line2D at 0x28b08cc4b20>]
```



In [40]:

```
X_final = X/H           # dtft for the  $X_{final}$ 
```

In [41]:

```
# using inverse fourier transform to convert X_final to X1:

N = len(X_final)
w = np.linspace(0,2*math.pi,N)          # taking 193 samples of w between 0 and 2*pi

X1 = np.zeros(N,dtype = 'complex_')

# Applying the summation formula for inverse dtft

for k in range(0,N-1):
    X1[k] = 0

    for n in range(0,N-1):

        X1[k] = ( X1[k] + X_final[n]* math.e**(1j * w[k] *n))

X1 = X1 / 193
```

Final plot for X1[n] and comparision with original x[n]

In [42]:

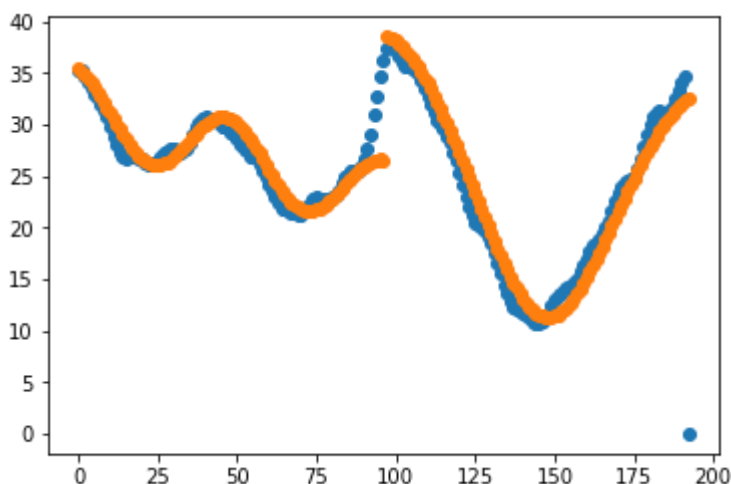
```
plt.scatter(range(0,193),abs(X1))
plt.scatter(range(0,193),x)
#plt.plot(w,y)

plt.show

# Orange plot represents original signal x[n]
# Blue plot represents X1[n]
```

Out[42]:

```
<function matplotlib.pyplot.show(close=None, block=None)>
```



In [43]:

```
# for comparing these plots we can calculate the mean error between all the values  
# As the value of error decreases, accuracy of determining x[n] increases.  
  
abs_error = []  
  
for i in range(0,191): # excluding last 2 values since they are  
    abs_error.append(abs(X1[i]+x[i]))  
  
mean_error = sum(abs_error)/191  
  
print(mean_error)
```

50.57979074862414

In []: