

DSP Homework

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Contents

1	Videos	1
2	How fast do Mosquitoes Flap	1
3	Base-band Wireless Communication Simulate	1
3.1	Simulation and Stability Analysis	1
3.2	Unstable Behavior When Noise is Zero	2
A	Comparison Between Code in Class and My Code	2
B	Code Listing	3

Abstract

1 Videos

2 How fast do Mosquitoes Flap

pass

3 Base-band Wireless Communication Simulate

3.1 Simulation and Stability Analysis

I wrote my code by imitating your code, but I made some changes because I thought there might be a problem with the code in class. This mistake will explain the unstable behavior when noise (along with calculation error) is zero. I wrote my comparison in appendix.

In recovery process, we have

$$\begin{aligned} y &= x * h + n \\ \tilde{y} &= \tilde{x} \times \tilde{h} + N_0 \end{aligned} \tag{1}$$

So, the recovered signal \hat{x} is as below

$$\begin{aligned} \mathcal{F}\{\hat{x}\} &= \tilde{y} \times \frac{1}{\tilde{h}} \\ &= (\tilde{x} \times \tilde{h} + N_0) \times \frac{1}{\tilde{h}} \\ &= \tilde{x} + \frac{N_0}{\tilde{h}} \end{aligned} \tag{2}$$

Here in (2), the IIR N_0/\tilde{h} , is why \hat{x} can be unstable.

3.2 Unstable Behavior When Noise is Zero

In mathematics, it is impossible to see unstable behavior in (3).

$$\begin{aligned}\mathcal{F}\{\hat{x}\} &= \tilde{y} \times \frac{1}{\tilde{h}} \\ &= (\tilde{x} \times \tilde{h}) \times \frac{1}{\tilde{h}} \\ &= \tilde{x}\end{aligned}\tag{3}$$

So, there must be other noise in the code. What I found is that there will be a very small error in division of python, and it is exactly this little error that makes $1/\tilde{h}$ to oscillate infinitely.

Please see the terminal output in Fig. 1 which proves my guess.

```
(xmh) xmh@xmh-PC:~/DSP$ /home/xmh/miniconda3/envs/xmh/bin/python /home/xmh/DSP/
xh[ 1 ] -= h[ 1 ] * xh[ 0 ], xh[ 0 ] = 1.0
xh[ 2 ] -= h[ 1 ] * xh[ 1 ], xh[ 1 ] = -0.0
xh[ 2 ] -= h[ 2 ] * xh[ 0 ], xh[ 0 ] = 1.0
xh[ 3 ] -= h[ 1 ] * xh[ 2 ], xh[ 2 ] = 2.9999999999999996
xh[ 3 ] -= h[ 2 ] * xh[ 1 ], xh[ 1 ] = -0.0
xh[ 4 ] -= h[ 1 ] * xh[ 3 ], xh[ 3 ] = -4.0000000000000002
xh[ 4 ] -= h[ 2 ] * xh[ 2 ], xh[ 2 ] = 2.9999999999999996
xh[ 5 ] -= h[ 1 ] * xh[ 4 ], xh[ 4 ] = 2.99999999999999947
xh[ 5 ] -= h[ 2 ] * xh[ 3 ], xh[ 3 ] = -4.0000000000000002
xh[ 6 ] -= h[ 1 ] * xh[ 5 ], xh[ 5 ] = -1.4401005572126297e-14
xh[ 6 ] -= h[ 2 ] * xh[ 4 ], xh[ 4 ] = 2.99999999999999947
xh[ 7 ] -= h[ 1 ] * xh[ 6 ], xh[ 6 ] = -3.0000000000000004
xh[ 7 ] -= h[ 2 ] * xh[ 5 ], xh[ 5 ] = -1.4401005572126297e-14
xh[ 8 ] -= h[ 1 ] * xh[ 7 ], xh[ 7 ] = 3.9999999999999999
xh[ 8 ] -= h[ 2 ] * xh[ 6 ], xh[ 6 ] = -3.0000000000000004
xh[ 9 ] -= h[ 1 ] * xh[ 8 ], xh[ 8 ] = 3.99999999999999954
xh[ 9 ] -= h[ 2 ] * xh[ 7 ], xh[ 7 ] = 3.9999999999999999
```

Figure 1: Error in Division Calculation

This error is revealed in float numbers like $2.99 \dots 96$ and $-4.00 \dots 02$ in line 4 and 6 of Fig. 1.

There are my comparison between whether there is noise or calculation error in signal y , see Fig. 2.

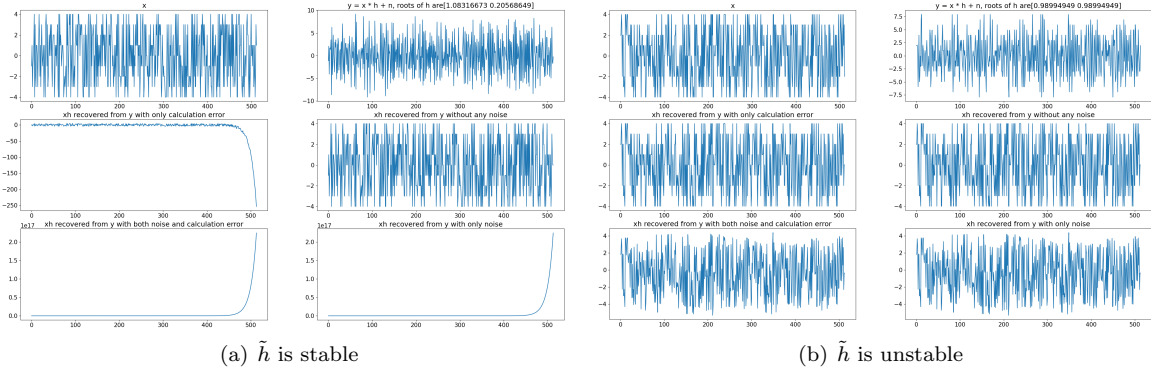


Figure 2: Stability Comparison when \tilde{h} is different

We can see in Fig. 2(a) that when \tilde{h} is unstable, \hat{x} recovered from y only with calculation error went infinite, but y without both calculation error and noise can be perfectly recovered. This phenomenon is because unstable amplified the noise or the calculation error in recovery, so if there's no disturb in y , it is possible to recover the original signal x .

Also, when \tilde{h} is stable or is decrement oscillation, the noise and the calculation error will be decreasing along with the recovery process, so without noise is no longer a necessary condition for signal recovery. However, in this condition, only y without any noise can be perfectly recovered ($\hat{x} = x$).

Appendix A Comparison Between Code in Class and My Code

Dear Yi, I may found a small mistake in your code on Tuesday's class. It is about the recovering of from $y = x * h$.

Please see the probably incorrect code as below:

```
def reX_estimate(y, h):
    N = len(y)
    M = len(h)
```

```

print("Length of y is ", N, ", Length of h is ", M)
xh = y.copy()
for n in range(N):
    print("Info: n is now ", n)
    for m in range(1, min(M, n)): # Mistake here, n should be n + 1
        xh[n] -= h[m] * xh[n - m]
        print("xh[" + n, "]" -= h[" + m, "]" * xh[" + n - m, "]")
    xh[n] /= h[0]
return xh

```

The length of h is 3 and the roots of h are about $[1.77, 0.57]$. Key point is in the output, see Fig. 3.

```

Length of y is 130 , Length of h is 3
Info: n is now 0
Info: n is now 1
Info: n is now 2
xh[ 2 ] -= h[ 1 ] * xh[ 1 ]
Info: n is now 3
xh[ 3 ] -= h[ 1 ] * xh[ 2 ]
xh[ 3 ] -= h[ 2 ] * xh[ 1 ]
Info: n is now 4
xh[ 4 ] -= h[ 1 ] * xh[ 3 ]
xh[ 4 ] -= h[ 2 ] * xh[ 2 ]

```

Figure 3: Output of Test Code

We can find that while $\hat{x}[2] = y[2] = h[0] \times x[2] + h[1] \times x[1] + h[2] \times x[0]$, this code only minus $h[1] \times x[1]$, that is the small mistake.

This small mistake in recovering x is *working as noise*, as a result, the $h(n)$ which is an unstable IIR, will *amplify this noise* and finally cause the *unstable behavior*. As the Fig. 4 revealed.

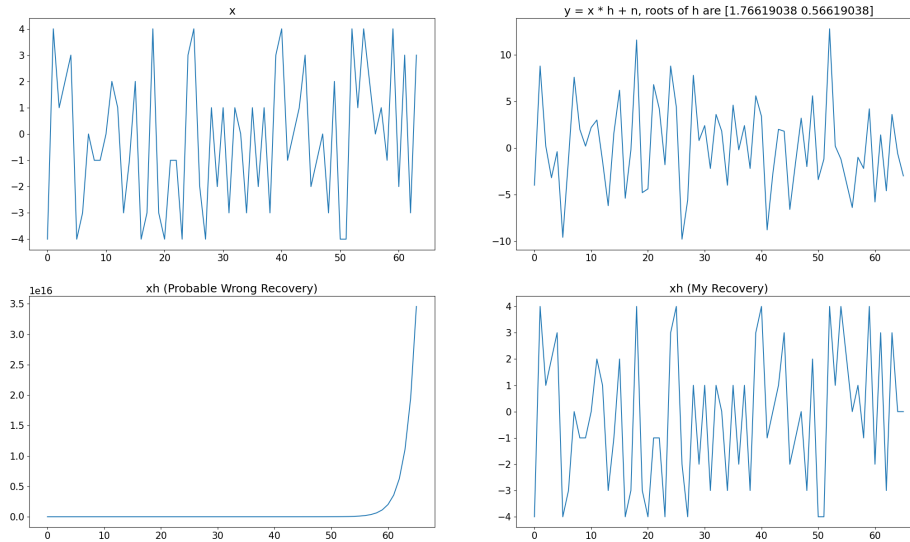


Figure 4: Different Behavior Between Previous Code and Current Code

Appendix B Code Listing

```

# simulate.py
import numpy as np
import matplotlib.pyplot as plt

```

```

plt.rcParams.update({'font.size': 15})

def genPAM(N, M):
    """Generate PAM Signal"""
    return (np.random.randint(-M, M + 1, (N)))

def genNoise(y, snr = 20):
    """Generate White Noise"""
    Es = np.mean(y**2)
    sigma = np.sqrt(Es / 10 ** ( snr / 10))
    n = sigma * np.random.randn(len(y))
    return n + y, n

def reX_estimate(y, h, ErrorFlag=1):
    """Estimate x From y"""
    N = len(y)
    M = len(h)
    xh = y.copy()
    for n in range(N):
        # print("Info: n is now ", n)
        for m in range(1, min(M, n + 1)):
            xh[n] -= h[m] * xh[n - m]
            # if ErrorFlag and n <= 20:
            #     print("xh[" , n, "] -= h[" , m, "] * xh[" , n - m, "]", " , xh[" , n - m, "] =", xh[n - m])
        xh[n] /= h[0]
        if ErrorFlag == 0:
            xh[n] = np.round(xh[n], 8) #Exclude Error generated by calculation accuracy
    return xh

#parameters
N = 256
Nc = 3
M = 4

#generate x and h randomly
x = genPAM(N, M)
h = np.random.randn(Nc)
# h = [1, 0, -0.98]

#generate y as received signal
y = np.convolve(x, h)
yn, n = genNoise(y, snr = 30)

#recover x in different methods
xh = reX_estimate(y, h)
xh_new = reX_estimate(y, h, ErrorFlag=0)
xhn = reX_estimate(yn, h)
xhn_new = reX_estimate(yn, h, ErrorFlag=0)

#draw the results and the comparison
fig = plt.figure(figsize=(12, 8))

ax = fig.add_subplot(3, 2, 1)
ax.plot(x)
ax.set_title('x')
ax = fig.add_subplot(3, 2, 2)
ax.set_title('y = x * h + n, roots of h are' + str(np.abs(np.roots(h))))
ax.plot(y)
ax = fig.add_subplot(3, 2, 3)
ax.set_title('xh recovered from y with only calculation error')
ax.plot(xh)
ax = fig.add_subplot(3, 2, 4)
ax.set_title('xh recovered from y without any noise')
ax.plot(xh_new)
ax = fig.add_subplot(3, 2, 5)
ax.set_title('xh recovered from y with both noise and calculation error')
ax.plot(xhn)
ax = fig.add_subplot(3, 2, 6)
ax.set_title('xh recovered from y with only noise')
ax.plot(xhn_new)

plt.show()

```