# CENG 384 - Signals and Systems for Computer Engineers Spring 2023 Homework 1

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1. (a) 
$$z = x + yj \longrightarrow \bar{z} = x - yj$$
  
 $2(x + yj) + 5 = j - (x - yj)$   
 $2x + 2yj + 5 = j - x + yj$   
 $3x + yj = -5 + j$   
 $x = \frac{-5}{3}, y = 1 \longrightarrow z = \frac{-5}{3} + j$   
 $|z|^2 = z\bar{z} = (\frac{-5}{3} + j)(\frac{-5}{3} - j)$   
 $= \frac{25}{9} + \frac{5j}{3} - \frac{5j}{3} - j^2 = \frac{25}{9} + 1 = \frac{34}{9}$ 

So,  $|z|^2 = \frac{34}{9}$  Also z can be plotted on the complex plane as below:

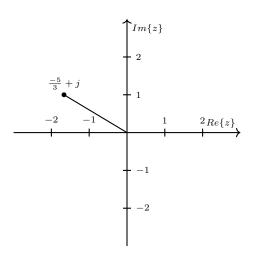


Figure 1:  $z = \frac{-5}{3} + j$  on the complex number

(b) 
$$z = re^{j\theta} \longrightarrow z^5 = r^5 e^{5j\theta}$$
  
 $z^5 = r^5 e^{5j\theta} = 32j$   
 $\implies r^5 = 32 \implies \mathbf{r} = \mathbf{2}$   
 $\implies e^{5j\theta} = j = e^{j\frac{\pi}{2}} \implies 5\theta = \frac{\pi}{2} + 2\pi k \implies \theta = \frac{\pi}{10} + \frac{\pi}{5}\mathbf{k}$   
Let's take k=0 where k $\in$ Z;  
 $\theta = \frac{\pi}{10}$ 

Thus,  $< z = \frac{\pi}{10}$  rad, |z| = 2 and its polar form is found as follows:

$$\mathbf{z} = 2\mathbf{e}^{\mathbf{j}\frac{\pi}{10}}$$

(c) Since the expressions includes multiplication and division of complex numbers, it is a wise choice to write subexpression in polar form:

$$z_1 = (1+j) = \sqrt{2}e^{j\frac{\pi}{4}}$$

$$z_2 = (\frac{1}{2} + \frac{\sqrt{3}}{2}j) = e^{j\frac{\pi}{3}}$$

$$z_3 = (j-1) = \sqrt{2}e^{j\frac{3\pi}{4}}$$

$$z = \frac{z_1 z_2}{z_3} = \frac{\sqrt{2}e^{j\frac{\pi}{4}}e^{j\frac{\pi}{3}}}{\sqrt{2}e^{j\frac{3\pi}{4}}} = e^{-j\frac{\pi}{6}}$$

So its magnitude |z|=1 and its angle  $< z = \frac{-\pi}{6} \text{ rad}{=}{-}30^{\circ}$ 

(d) It is known that  $j = e^{j\frac{\pi}{2}}$ 

$$z = je^{-j\frac{\pi}{2}} = e^{j\frac{\pi}{2}}e^{-j\frac{\pi}{2}} = e^0 = \mathbf{1}$$

We know ,  $1 = e^{j2\pi}$ 

So ,  $z=e^{j2\pi}$ , which is in polar form.

2. x(t) can be defined in the form of a piecewise function as follows:

$$x(t) = \begin{cases} 0, & \text{if } -3 \le t \le -2\\ t+2, & \text{if } -2 < t \le -1\\ 1, & \text{if } -1 < t \le 1\\ -t+2, & \text{if } 1 < t \le 2\\ 0, & \text{if } 2 < t \le 3 \end{cases}$$

Then, replace t with  $\frac{1}{2}t + 1$  in the function above:

$$x(\frac{1}{2}t+1) = \begin{cases} 0, & \text{if } -3 \le \frac{1}{2}t+1 \le -2\\ \frac{1}{2}t+3, & \text{if } -2 < \frac{1}{2}t+1 \le -1\\ 1, & \text{if } -1 < \frac{1}{2}t+1 \le 1\\ -\frac{1}{2}t+1, & \text{if } 1 < \frac{1}{2}t+1 \le 2\\ 0, & \text{if } 2 < \frac{1}{2}t+1 \le 3 \end{cases}$$

This is equivalent to:

$$x(\frac{1}{2}t+1) = \begin{cases} 0, & \text{if } -8 \le t \le -6\\ \frac{1}{2}t+3, & \text{if } -6 < t \le -4\\ 1, & \text{if } -4 < t \le 0\\ -\frac{1}{2}t+1, & \text{if } 0 < t \le 2\\ 0, & \text{if } 2 < t \le 4 \end{cases}$$

In a more intuitive way, transformation applied to x(t) can be described as:

- (i) Time scale: expanded by 2
- (ii) Time shift: Shift by 2 to the left

Thus the signal  $y(t) = x(\frac{1}{2}t + 1)$  can be drawn as:

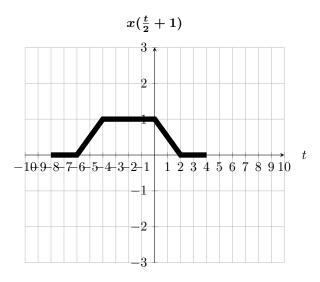


Figure 2:  $t \text{ vs. } y(t) = x(\frac{t}{2} + 1).$ 

3. (a) x[-n]: It is the reflected version of x[n] about y-axis.

x[2n-1]:To get this, we first shrink x[n] by 2, then shift to right by 1/2 and only take the integer n values, since it is discrete time signal.

At the end, we sum up these two signal, which are x[-n] and x[2n-1].

To be more precise:

<u>n</u>	$\mathbf{x}[\mathbf{n}]$		<u>n</u>	$\mathbf{x}[-]$
-1	0		<u>n</u> -8 -7	0
0	0		-7	3
1	-1		-6	0
2	2		-5	0
n -1 0 1 2 3 4 5 6 7 8	0		-6 -5 -4 -3 -2 -1 0	-4
4	-4		-3	0
5	0		-2	$\begin{array}{c c} 0 \\ 2 \end{array}$
6	0		-1	-1
7	$0 \\ 3$		0	0
8	0		1	0
		'		

n	x[2n-1]
0	0
1	-1
2 3	0
3	0
4	3

Therefore;

<u>n</u>	$\mathbf{x}[-\mathbf{n}] + \mathbf{x}[2\mathbf{n} - 1]$
-8	0
n   -8   -7   -6   -5   -4   -3   -2   -1   0	$egin{pmatrix} 0 \\ 3 \\ 0 \\ \end{bmatrix}$
-6	0
-5	0
-4	-4
-3	$egin{pmatrix} 0 \ 2 \end{bmatrix}$
-2	2
-1	-1
0	0
1	-1
1 2 3 4	0
3	0
4	3

So x[-n]+x[2n-1] can be drawn as:

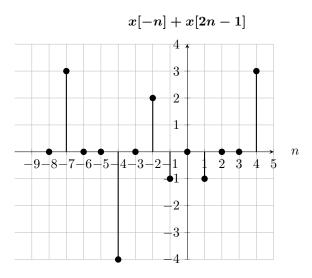


Figure 3: n vs. x[-n] + x[2n-1].

(b) 
$$x[-n] + x[2n-1] = 3\delta[n+7] - 4\delta[n+4] + 2\delta[n+2] - \delta[n+1] - \delta[n-1] + 3\delta[n-4]$$

## 4. (a) YES. It is periodic.

$$x(t) = 5sin(3t - \frac{\pi}{4})$$

$$w_0 = 3 \longrightarrow T_0 = \frac{2\pi}{3}$$

Therefore, it is periodic with fundamental period  $T_0 = \frac{2\pi}{3}$ 

(b) YES. It is periodic.

Let's say 
$$x[n] = x_1[n] + x_2[n]$$

$$x_1[n] = cos(\frac{13\pi}{10}n)$$
 and  $x_2 = sin(\frac{7\pi}{10}n)$ .

For  $x_1$ ;

$$w_0 = \frac{13\pi}{10} \longrightarrow N_0 = \frac{2\pi}{w_0} m = \frac{2\pi}{13\frac{\pi}{10}} m = \frac{20}{13} m.$$

Put 13 into m 
$$\longrightarrow N_0 = \frac{20}{13}13 = 20$$

For  $x_2$ ;

$$w_0 = \frac{7\pi}{10} \longrightarrow N_0 = \frac{2\pi}{w_0} m = \frac{2\pi}{7\frac{\pi}{10}} m = \frac{20}{7} m.$$

Put 7 into m 
$$\longrightarrow N_0 = \frac{20}{7}7 = 20$$

We need to find lowest common multiple of fundamental periods of  $x_1$  and  $x_2$  to find fundamental period of x[n]:

$$LCM(20,20)=20$$

So, this signal, x[n], is periodic with fundamental period  $N_0 = 20$ 

(c) NO. It is not periodic.

$$w_0 = 7 \longrightarrow N_0 = \frac{2\pi}{w_0} m = \frac{2\pi}{7} m$$

There is no integer value of m which makes  $N_0$  an integer.

We know that fundamental period of a discrete time signal must be an integer. However, we've seen that there is no integer value of m making  $N_0$  integer.

Thus, this signal is not periodic.

- 5. (a) x(t)=u(t-1)-3u(t-3)+u(t-4)
  - (b) We know that  $\frac{du(t-t_0)}{dt} = \delta(t-t_0)$

By using this fact,  $\frac{dx(t)}{dt}$  can be found as follows:

$$\frac{dx(t)}{dt} = \delta(t-1) - 3\delta(t-3) + \delta(t-4).$$

It is drawn as below:

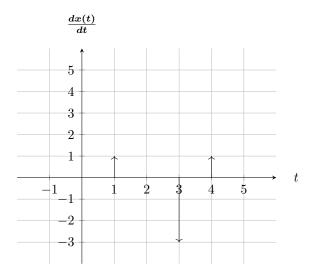


Figure 4: t vs.  $\frac{dx(t)}{dt}$ .

6. (a) Memory:YES. It has memory. It depends on future or past value of the input e.g.

$$y(2)=2x(7)$$

$$y(-5) = -5x(-7)$$

**Stability:** NO. It is not stable. Even if input, x(t), would be bounded y(t) may become unbounded because of the term, t, which is unbounded. So, y(t) is unstable.

Causality: NO. It is not casual. The outputs depends on future values of input. For example;

$$y(2) = 2x(7)$$

### **Linearity**:YES. It is linear.

Suppose we feed two inputs,  $x_1$  and  $x_2$  to the system. The corresponding outputs will be:

$$x_1(t) \longrightarrow y_1(t) = tx_1(2t+3)$$
  
 $x_2(t) \longrightarrow y_2(t) = tx_2(2t+3)$ 

Output for the superposition of the inputs,  $x_3 = a_1x_1 + a_2x_2$  is:

$$y_3(t) = t(a_1x_1 + a_2x_2) = a_1y_1 + a_2y_2$$

Superposition holds. Thus it is linear.

Invertibility: NO. It is not invertible.

$$x(t) = \frac{y(\frac{t-3}{2})}{\frac{t-3}{2}}$$
, not defined when t=3.

time – invariance: *NO*. It is time-varying(Not time invariant).

First, shift the input by  $t_0$  and observe the corresponding output.

Corresponding output:  $tx(2(t-t_0)+3) = tx(2t-2t_0+3)$ .

Then, shift output by  $t_0$ :

$$(t - t_0)x(2(t - t_0) + 3) = (t - t_0)x(2t - 2t_0 + 3)$$

Lastly, observe that

$$tx(2t-2t_0+3) \neq (t-t_0)x(2t-2t_0+3)$$

Thus, it is time-varying.

(b) Memory: YES. It has memory. The output depends on past input values. For example;

$$y[0]=x[-1]+x[-2]+x[-3]+...$$

**Stability:** NO. It is not stable. Even if the input is bounded, the output is infinite sum of the input, i.e. y[n] is the sum of infinitely many number of bounded values. Therefore, the system is not stable.

Causality: YES. It is causal. The output does not depend on future values of the input. It only depends on past values of input. For example:

$$y[n]=x[n-1]+x[n-2]+....$$

So, it is causal.

**Linearity**:YES. It is linear. Suppose , we feed two inputs,  $x_1$  and  $x_2$ , to the system. The corresponding outputs will be:

$$x_1[n] \longrightarrow y_1[n] = \sum_{k=1}^{\infty} ax_1[n-k]$$

$$x_2[n] \longrightarrow y_2[n] = \sum_{k=1}^{\infty} ax_2[n-k]$$

Output for the superposition of the inputs,

$$x_3 = ax_1 + ax_2$$

$$x_3[n] \longrightarrow y_3[n] = \sum_{k=1}^{\infty} a_1 x_1[n-k] + a_2 x_2[n-k]$$

$$= a_1 \sum_{k=1}^{\infty} x_1[n-k] + a_2 \sum_{k=1}^{\infty} x_1[n-k]$$

$$= a_1 y_1[n] + a_2 y_2[n]$$

Superposition holds. thus it is linear

#### Invertibility: YES. It is invertible.

$$y[n+1] = x[n] + x[n-1] + x[n-2] + \dots$$
  
 $y[n] = x[n-1] + x[n-2] + x[n-3] + \dots$ 

If we substract y[n+1] and y[n]:

$$y[n+1] - y[n] = x[n]$$

We found x[n] = y[n+1] - y[n]. Thus the system is invertible.

### time - invariance: YES. It is time-invariant.

First, shift the input by  $n_0$  and observe the corresponding output:

$$\sum_{k=1}^{\infty} x[n - n_0 - k]$$

Then, shift the output by  $n_0$ :

$$\sum_{k=1}^{\infty} x[n - n_0 - k]$$

As you can see, both of them are equal. A time shift in input creates identical shift on the output.

Thus it is time-invariant.

```
7. (a) # Question 7 - Part a
     2 import matplotlib.pyplot as plt
     _4 MARGIN = 0
     6 def load_data_part_a(filepath):
          file = open(filepath)
          data =list(map(lambda x: float(x), file.readline().split(",")))
     9
           starting_index = data[0]
          signal = data[1:]
    10
          time = list(range(int(starting_index), int(starting_index) + len(signal)))
    11
          temp = fill_beyond_with_0(time, signal)
    12
          time = temp[0]
    13
          signal = temp[1]
          return (time, signal)
    15
    16
    17 def fill_beyond_with_0(time, signal):
         zeros = [0] * MARGIN
    18
           new_signal = zeros + signal + zeros
    19
          lower_bound = time[0]
    20
          upper_bound = time[-1]
    21
          new_lower_bound = lower_bound - MARGIN
          new_upper_bound = upper_bound + MARGIN
    23
          new_time = list(range(new_lower_bound, lower_bound)) + time + list(range(upper_bound+1,
    24
          new_upper_bound+1))
          return (new_time, new_signal)
    25
    27 def reflect_about_y(time, signal):
         reflected_signal = signal[::-1]
    28
          reflected_time = list(map(lambda x: -x, time[::-1]))
          return (reflected_time, reflected_signal)
    30
    32 def magnitude_scale(time, signal, k):
```

```
33
     scaled_signal = []
     for element in signal:
34
35
           scaled_signal.append(k * element)
      return (time, scaled_signal)
37
40 \# x(t) = [0.5 * (x(t) + x(-t))] + [0.5 * (x(t) - x(-t))]
41 # even part: [0.5 * (x(t) + x(-t))]
42 \text{ # odd part: } [0.5 * (x(t) - x(-t))]
43
^{44} def even_odd_decomposition(filepath, option): # option = 0 -> even part , option = 1 -> odd part
       x_n = load_data_part_a(filepath) # x[n]
45
      x_minus_n = reflect_about_y(x_n[0], x_n[1]) # x[-n]
46
47
48
      x_n_{time} = x_n[0]
      x_minus_n_time = x_minus_n[0]
50
51
      x_n_{signal} = x_n[1]
      x_minus_n_signal = x_minus_n[1]
53
       x_n_lower_time_bound = x_n_time[0]
54
      x_n_upper_time_bound = x_n_time[-1]
55
56
       x_minus_n_lower_time_bound = x_minus_n_time[0]
57
      x_minus_n_upper_time_bound = x_minus_n_time[-1]
58
59
60
       new_time_lower_bound = min(x_n_lower_time_bound, x_minus_n_lower_time_bound)
      new_time_upper_bound = max(x_n_upper_time_bound, x_minus_n_upper_time_bound)
61
62
       i = new_time_lower_bound
63
64
      tmp1 = []
       c1 = 0
65
       while (i < x_n_lower_time_bound):</pre>
66
67
          c1 += 1
           tmp1.append(i)
          i+=1
69
70
     i = x_n_upper_time_bound
71
      tmp2 = []
72
73
      c2 = 0
      while (i < new_time_upper_bound):</pre>
74
          c2+=1
75
           tmp2.append(i)
          i+=1
77
78
79
      x_n_{time} = tmp1 + x_n_{time} + tmp2
      x_n_{signal} = c1 * [0] + x_n_{signal} + c2 * [0]
80
81
82
      i = new_time_lower_bound
83
      tmp1 = []
      c1 = 0
85
       while (i < x_minus_n_lower_time_bound):</pre>
86
          c1+=1
87
88
           tmp1.append(i)
          i += 1
89
90
      i = x_minus_n_upper_time_bound
91
       tmp2 = []
      c2 = 0
93
       while (i < new_time_upper_bound):</pre>
94
           c2+=1
95
           tmp2.append(i)
96
           i += 1
97
98
       x_minus_n_time = tmp1 + x_minus_n_time + tmp2
99
      x_minus_n_signal = c1 * [0] + x_minus_n_signal + c2 * [0]
      if (option == 0):
103
           even_decomposition_time = x_n_time
104
105
           even_decomposition_signal = []
106
           for i in range(len(even_decomposition_time)):
107
               even_decomposition_signal.append(x_n_signal[i] + x_minus_n_signal[i])
108
109
           return magnitude_scale(even_decomposition_time, even_decomposition_signal, 0.5)
110
111
       elif (option == 1):
112
           odd_decomposition_time = x_n_time
114
           odd_decomposition_signal = []
```

```
for i in range(len(odd_decomposition_time)):
116
               odd_decomposition_signal.append(x_n_signal[i] - x_minus_n_signal[i])
117
118
          return magnitude_scale(odd_decomposition_time, odd_decomposition_signal, 0.5)
119
120
121
123
124
125 ############## PART A - Outputs ################
126 r = even_odd_decomposition("./chirp_part_a.csv", 0)
127 plt.stem(r[0], r[1])
128 plt.xlabel('n', fontsize=20)
plt.ylabel('Even(x[n])', fontsize=20)
130 plt.title('Even Part of chirp_part_a.csv', fontsize=20)
131 plt.savefig("Even Part of chirp_part_a.pdf", format="pdf", bbox_inches="tight")
132 plt.cla()
133 r = even_odd_decomposition("./chirp_part_a.csv", 1)
134 plt.stem(r[0], r[1])
135 plt.xlabel('n', fontsize=20)
136 plt.ylabel('0dd(x[n])', fontsize=20)
137 plt.title('Odd Part of chirp_part_a.csv', fontsize=20)
138 plt.savefig("Odd Part of chirp_part_a.pdf", format="pdf", bbox_inches="tight")
139 plt.cla()
140 r = even_odd_decomposition("./shifted_sawtooth_part_a.csv", 0)
141 plt.stem(r[0], r[1])
142 plt.xlabel('n', fontsize=20)
plt.ylabel('Even(x[n])', fontsize=20)
144 plt.title('Even Part of shifted_sawtooth_part_a.csv', fontsize=20)
145 plt.savefig("Even Part of shifted_sawtooth_part_a.pdf", format="pdf", bbox_inches="tight")
146 plt.cla()
147 r = even_odd_decomposition("./shifted_sawtooth_part_a.csv", 1)
148 plt.stem(r[0], r[1])
149 plt.xlabel('n', fontsize=20)
150 plt.ylabel('Odd(x[n])', fontsize=20)
plt.title('Odd Part of shifted_sawtooth_part_a.csv', fontsize=20)
152 plt.savefig("Odd Part of shifted_sawtooth_part_a.pdf", format="pdf", bbox_inches="tight")
153 plt.cla()
154
155 r = even_odd_decomposition("./sine_part_a.csv", 0)
156 plt.stem(r[0], r[1])
plt.xlabel('n', fontsize=20)
158 plt.ylabel('Even(x[n])', fontsize=20)
159 plt.title('Even Part of sine_part_a.csv', fontsize=20)
160 plt.savefig("Even Part of sine_part_a.pdf", format="pdf", bbox_inches="tight")
161 plt.cla()
162 r = even_odd_decomposition("./sine_part_a.csv", 1)
163 plt.stem(r[0], r[1])
164 plt.xlabel('n', fontsize=20)
165 plt.ylabel('Odd(x[n])', fontsize=20)
166 plt.title('Odd Part of sine_part_a.csv', fontsize=20)
167 plt.savefig("Odd Part of sine_part_a.pdf", format="pdf", bbox_inches="tight")
168 plt.cla()
```

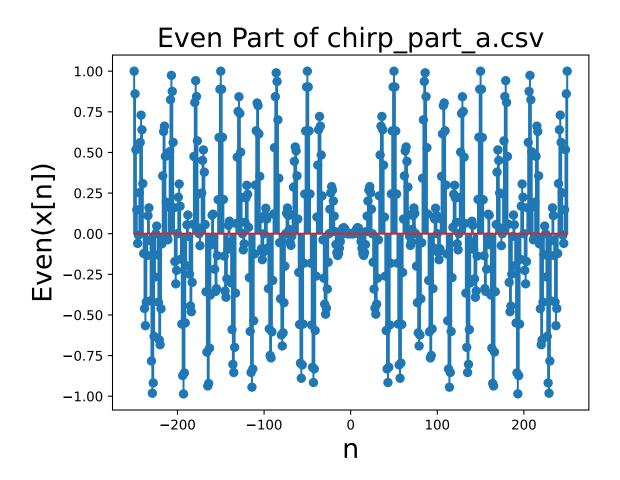


Figure 5: Even Part of chirp\_part\_a.csv

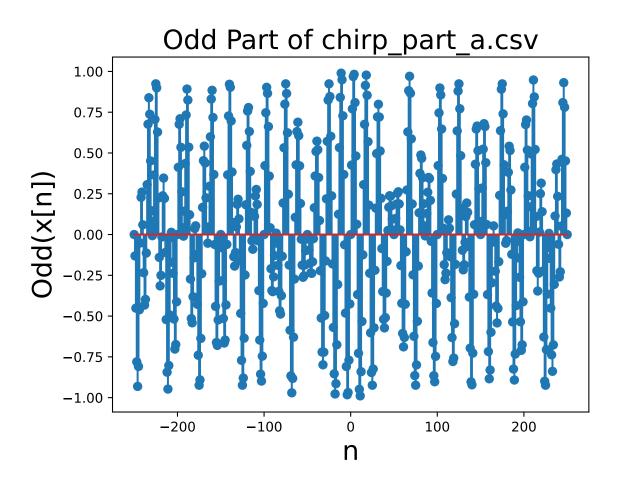


Figure 6: Odd Part of chirp\_part\_a.csv

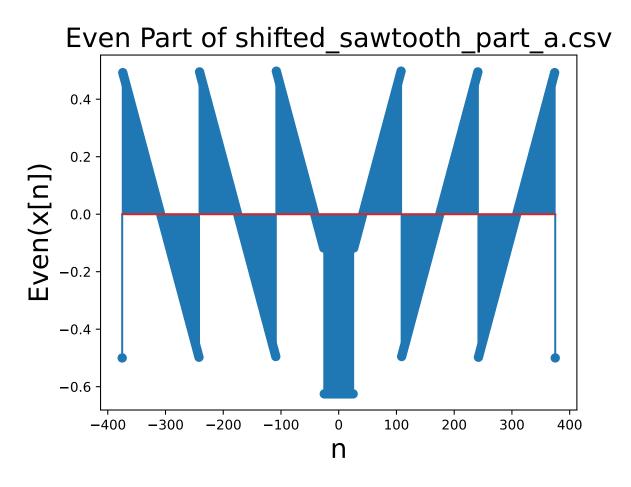


Figure 7: Even Part of shifted\_sawtooth\_part\_a.csv

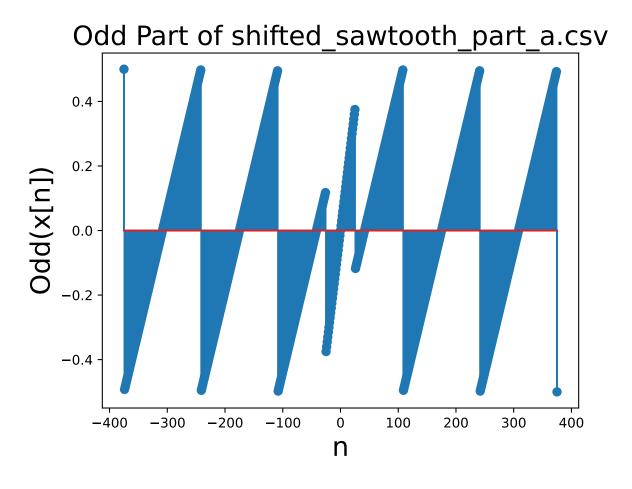


Figure 8: Odd Part of shifted\_sawtooth\_part\_a.csv

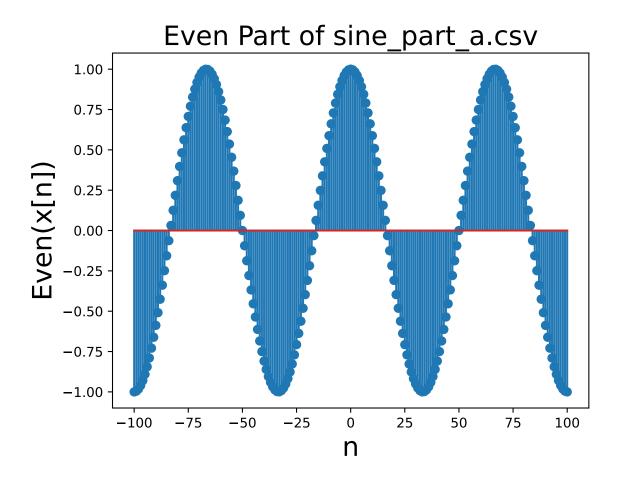


Figure 9: Even Part of sine\_part\_a.csv

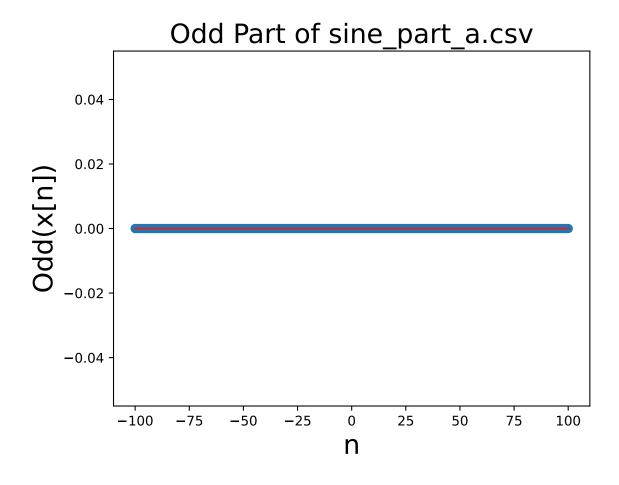


Figure 10: Odd Part of sine\_part\_a.csv

```
(b) # Question 7 - Part b
 2 import matplotlib.pyplot as plt
 4 MARGIN = O
 6 def load_data_part_b(filepath):
       file = open(filepath)
       data =list(map(lambda x: float(x), file.readline().split(",")))
       starting_index = data[0]
      a = data[1]
 10
 11
      b = data[2]
       signal = data[3:]
       time = list(range(int(starting_index), int(starting_index) + len(signal)))
 13
       temp = fill_beyond_with_0(time, signal)
 14
       time = temp[0]
 15
       signal = temp[1]
 16
      return (time, signal, a, b)
 17
 18
 19 def fill_beyond_with_0(time, signal):
       zeros = [0] * MARGIN
       new_signal = zeros + signal + zeros
 21
       lower_bound = time[0]
 22
       upper_bound = time[-1]
 23
       new_lower_bound = lower_bound - MARGIN
 24
       new_upper_bound = upper_bound + MARGIN
       new_time = list(range(new_lower_bound, lower_bound)) + time + list(range(upper_bound+1,
 26
       new_upper_bound+1))
       return (new_time, new_signal)
 29 def reflect_about_y(time, signal):
 30
       reflected_signal = signal[::-1]
       reflected_time = list(map(lambda x: -x, time[::-1]))
 31
       return (reflected_time, reflected_signal)
 32
 33
 34 def time_shift(time, signal, amount, direction): # direction : 0 -> left shift , direction : 1
       -> right shift
       shifted_time = []
 35
       if (direction == 0):
 36
           for moment in time:
 37
               shifted_moment = moment - amount
 38
 39
               shifted_time.append(shifted_moment)
       else:
 40
 41
           for moment in time:
                shifted_moment = moment + amount
               shifted_time.append(shifted_moment)
 43
 44
       return (shifted_time, signal)
 45
 46
 47 def time_scale(time, signal, a):
     if (a == 1):
 48
           return (time, signal)
 49
       elif (a == -1):
 51
           return reflect_about_y(time, signal)
       elif (a > 0):
 52
          scaled_time = []
 53
           scaled_signal = []
 54
 55
           for i in range(0,len(time)):
               if (int(time[i] / a) == float(time[i] / a)):
 56
                   scaled_time.append(int(time[i] / a))
 57
                    scaled_signal.append(signal[i])
           return (scaled_time, scaled_signal)
 59
 60
      elif (a < 0):</pre>
           reflected_version = reflect_about_y(time, signal)
 61
           time = reflected_version[0]
 62
           signal = reflected_version[1]
 63
 64
           a = -a
           scaled_time = []
 65
           scaled_signal = []
           for i in range(0,len(time)):
 67
                if (int(time[i] / a) == float(time[i] / a)):
 68
                   scaled_time.append(int(time[i] / a))
                   scaled_signal.append(signal[i])
 70
           return (scaled_time, scaled_signal)
 71
 74 def x_an_b(filepath):
       data = load_data_part_b(filepath)
 75
       time = data[0]
 76
       signal = data[1]
 77
      a = data[2]
 78
      b = data[3]
 79
       if (b == 0):
```

```
return time_scale(time, signal, a)
      elif (b > 0):
82
83
          shifted_version = time_shift(time, signal, abs(b), 0)
          return time_scale(shifted_version[0], shifted_version[1], a)
      elif (b < 0):</pre>
85
86
          shifted_version = time_shift(time, signal, abs(b), 1)
          return time_scale(shifted_version[0], shifted_version[1], a)
90 ############# PART B - Outputs ################
91 r = x_an_b("./chirp_part_b.csv")
92 plt.stem(r[0], r[1])
93 plt.xlabel('n', fontsize=20)
94 plt.ylabel('x[an + b]', fontsize=20)
95 plt.title("chirp_part_b.csv", fontsize=20)
96 plt.savefig("chirp_part_b.pdf", format="pdf", bbox_inches="tight")
98 r = x_an_b("./shifted_sawtooth_part_b.csv")
99 plt.stem(r[0], r[1])
100 plt.xlabel('n', fontsize=20)
plt.ylabel('x[an + b]', fontsize=20)
102 plt.title("shifted_sawtooth_part_b.csv", fontsize=20)
103 plt.savefig("shifted_sawtooth_part_b.pdf", format="pdf", bbox_inches="tight")
104
105 r = x_an_b("./sine_part_b.csv")
106 plt.stem(r[0], r[1])
107 plt.xlabel('n', fontsize=20)
108 plt.ylabel('x[an + b]', fontsize=20)
109 plt.title("sine_part_b.csv", fontsize=20)
110 plt.savefig("sine_part_b.pdf", format="pdf", bbox_inches="tight")
```

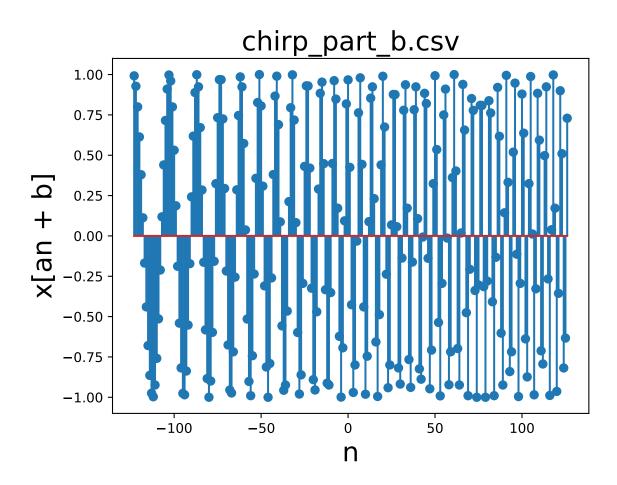


Figure 11: x[an+b] corresponding to chirp\_part\_b.csv

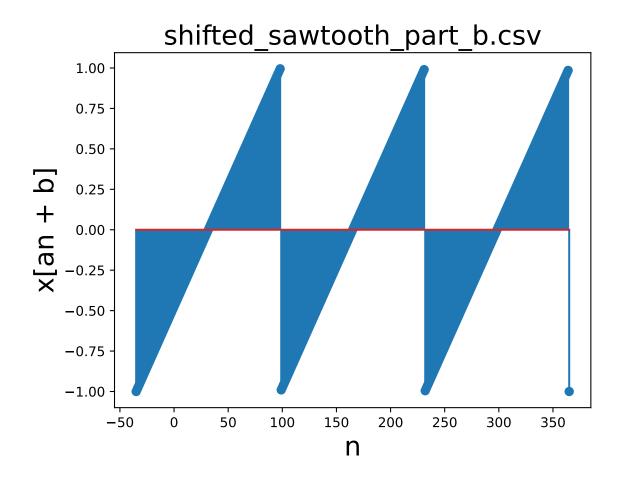


Figure 12: x[an+b] corresponding to shifted\_sawtooth\_part\_b.csv

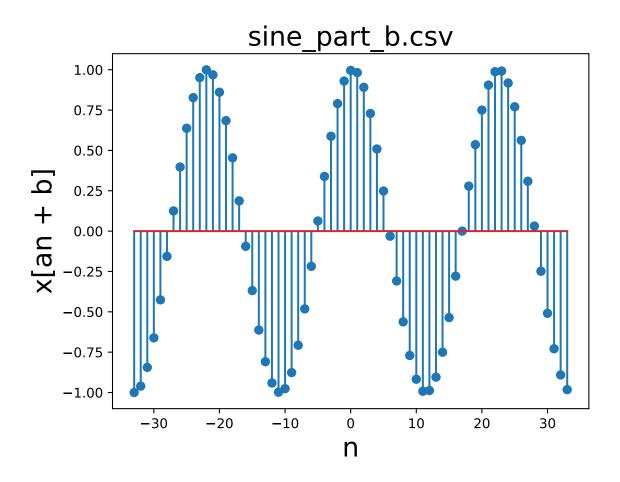


Figure 13: x[an+b] corresponding to sine\_part\_b.csv