

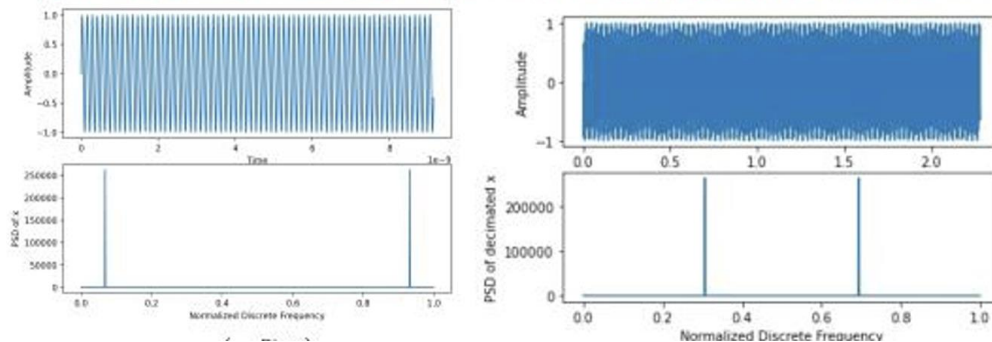
ECEN 610: MIXED SIGNAL INTERFACE HW.1

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How to find the Alias Frequency F_a if Decimation is used?



HW: Find a methodology that finds F_{in} such that the alias frequency before and after decimation are in the same discrete frequency and the decimated alias frequency falls on a discrete frequency bin

- **Decimation** is the process of **reducing the sampling rate F_s** of a digital signal.
- When a signal is downsampled (Decimation), its frequency components may exceed the new Nyquist frequency:

$$F'_{nyquist} = F_s / 2M \quad | \quad M = \text{decimate factor}$$

This causes high-frequency components to **fold back (alias) into the Nyquist range**, resulting in aliasing.

$$F_s D = F_s / M$$

$$\text{Nyquist } D = F_s D / 2$$

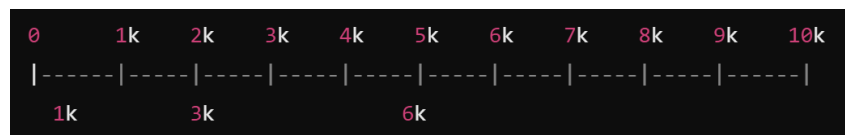
Ex: $F_s = 10k$ $F_{in} = 3k$ and $6k$ Nyquist BW = $5k$

3k no alias, but 6k alias back to 4k

After decimation $M=2$ $F_s D = 5k$ Nyquist BW = $2.5k$

Now 3k alias back to 2k, and (6k alias to 4k) and then back to 1k

Original:



Decimate:



- Bin: When we perform Fast Fourier Transform (FFT) on the signal, the frequency axis will be divided into N bins, each bin represents a fixed frequency range.

Ex: $F_s=10\text{k Hz}$ FFT N :1024

Bin: $\Delta f= 10\text{k}/1024= 9.77\text{Hz}$

Bin1: 0Hz Bin2: 9.77Hz Bin3: 19.54Hz ...Bin1024

- FFT can only resolve discrete frequency points. These frequency points are determined by FFT bins(No matter before decimating or after). The bin intervals are:

Bin size= F_s/N

If alias frequency f_a does not fall exactly on a certain bin, then it will have a frequency error in the FFT spectrum.

Therefore, we use:

*$F_a= \text{round}(F_a/\text{Bin size}) * \text{Bin size}$*

Align to the nearest FFT bin so that the FFT can parse it correctly.

Ex: $F_s= 1000\text{Hz}$, FFT N: 16, $M=2$ ($F_sD=500\text{Hz}$ $ND=8$), $F_{in}:1200\text{Hz}$

Original F_a : $|1200- 1*1000|= 200\text{Hz}$

FFT Bin: $F_s/N= 62.5\text{Hz}$

Normalized F_a : $\text{around}(F_a/\text{Bin size}) * \text{bin size}= 187.5\text{Hz}$ (Bin3)

$F_aD= F_a \bmod F_sD= 200 \bmod 500= 200\text{Hz}$

Still need to normalized again.

■ Math for doing this

1. $F_a=|F_{in}-kF_s|$, $k=\text{round}(f_{in}/f_s)$
 2. $F_aD= F_a \bmod F_sD$
 3. Normalized Bin: $\text{Bin}=\text{round}(F_a/\text{Bin size}) * \text{Bin size}$ (for both F_a F_aD)
(F_a normalize and F_aD normalized should be the same)
- ✧ After downclocking, the FFT Bin spacing (frequency resolution) remains unchanged.
 - ✧ After downclocking, the number of FFT Bins is reduced, but the frequency point spacing remains the same.
 - ✧ If alias frequency F_a corresponding bin remains unchanged, it will still fall into the same frequency bin after downscaling to ensure correct alignment.

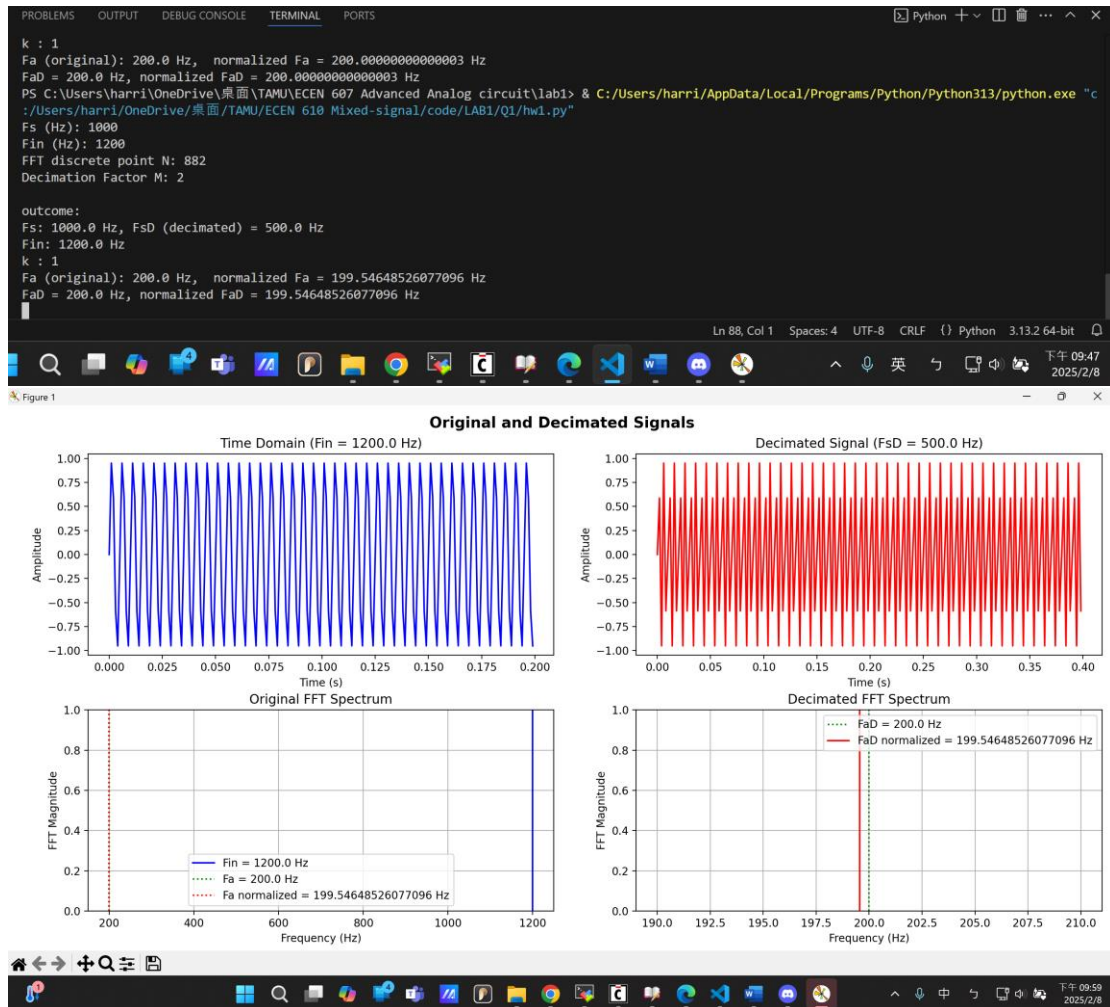
● Simulation

$F_s= 1000\text{Hz}$, $F_{in}: 1200\text{Hz}$, FFT N:882, $M=2$

Estimation: $f_a=|f_{in}-kF_s|=200$, $k=\text{round}(f_{in}/f_s)=1$

Bin size: $F_s/N= 1000/882= 1.13379\text{Hz}$

Normalized $F_a= \text{round}(F_s/ \text{Bin size}) * \text{Bin sized}= 199.5468\text{Hz}$



Code for Fa and normalized

```

C: > Users > harri > OneDrive > 桌面 > TAMU > ECEN 610 Mixed-signal > code > LAB1 > Q1 > hw1.py > ...
1 import numpy as np
2 import matplotlib.pyplot as plt
3 from scipy.fftpack import fft
4
5 Fs = float(input("Fs (Hz): "))
6 Fin = float(input("Fin (Hz): "))
7 N = int(input("FFT discrete point N: ")) # EX: N=100 fin(Hz) cut into x[1-100] discrete point x[y]= fin/N * y
8 M = int(input("Decimation Factor M: ")) # Decimation Factor EX: Fs/M into FsD
9
10 # After decimate FsD and ND
11 FsD = Fs / M
12 ND = N // M # 降頻後的 FFT 點數
13
14 # Fa (original)
15 k = round(Fin / Fs) # closest kFs
16 Fa = abs(Fin - k * Fs) # Frequene alias back to Nyquist BW
17
18 # FaD (mod to insure between FsD)
19 FaD = Fa % FsD
20
21 # bin size
22 bin_size = Fs / N
23 bin_size_D = FsD / ND
24
25 # Fix alias freq: fit FFT bin
26 normalized_Fa = round(Fa / bin_size) * bin_size
27 normalized_FaD = round(FaD / bin_size_D) * bin_size_D

```

FFT and pre-plotting

```

30
31 # time
32 t = np.arange(N) / Fs # original time
33 t_decimated = np.arange(ND) / FsD # D time
34
35 # Original Fin
36 x_original = np.sin(2 * np.pi * Fin * t)
37 x_alias = np.sin(2 * np.pi * normalized_Fa * t) # 修正後 alias 頻率
38
39 # decimated freq
40 x_decimated = x_original[::M] # FsD= Fs/M
41
42 # FFT [:N//2]slicing
43 X_original = np.abs(fft(x_original)) # fft abs conjurate for Magnitude not Phase
44 X_decimated = np.abs(fft(x_decimated)) # fft output [-Fs/2- Fs/2] only keep the 0- Fs/2
45
46 # fft (0, Fs/N, Fs/2N ... {N/(2-1)Fs}/N, -Fs/2 )
47 # [:N//2] for the first half ( to show nyquist BW)
48 # [N//2:] for the second half ( negative part )
49
50 freqs = np.fft.fftfreq(N, d=1/Fs) # FFT freq
51 freqs_decimated = np.fft.fftfreq(ND, d=1/FsD) # D FFT freq

```

Plotting

```

54
55 # Print
56 print("\noutcome:")
57 print(f"Fs: {Fs} Hz, FsD (decimated) = {FsD} Hz")
58 print(f"Fin: {Fin} Hz")
59 print(f"k : {k}")
60 print(f"Fa (original): {Fa} Hz, normalized Fa = {normalized_Fa} Hz")
61 print(f"FaD = {FaD} Hz, normalized FaD = {normalized_FaD} Hz")
62
63 # Plot
64 fig, axes = plt.subplots(2, 2, figsize=(12, 8))
65
66 # Upper left: original signal (time domain)
67 axes[0, 0].plot(t[:200], x_original[:200], color='blue')
68 axes[0, 0].set_title(f"Time Domain (Fin = {Fin} Hz)")
69 axes[0, 0].set_xlabel("Time (s)")
70 axes[0, 0].set_ylabel("Amplitude")
71
72 # Upper right: signal after downconversion (time domain)
73 axes[0, 1].plot(t_decimated[:200], x_decimated[:200], color='red')
74 axes[0, 1].set_title(f"Decimated Signal (Fs' = {FsD} Hz)")
75 axes[0, 1].set_xlabel("Time (s)")
76 axes[0, 1].set_ylabel("Amplitude")
77
78 # Bottom left: original FFT spectrum
79 axes[1, 0].plot(freqs, X_original, label=f"Fin = {Fin} Hz", color='blue')
80 axes[1, 0].axvline(Fin, color='blue', label=f"Fin = {Fin} Hz") # 原始頻率標示
81 axes[1, 0].axvline(Fa, color='green', linestyle=":", label=f"Fa = {Fa} Hz")
82 axes[1, 0].axvline(normalized_Fa, color='red', linestyle=":", label=f"Fa normalized = {normalized_Fa} Hz") # Alias 頻率標示
83 axes[1, 0].set_xlabel("Frequency (Hz)")
84 axes[1, 0].set_ylabel("FFT Magnitude")
85 axes[1, 0].set_title("Original FFT Spectrum")
86 axes[1, 0].legend()
87
88
89 # Bottom right: FFT spectrum after downconversion
90 axes[1, 1].plot(freqs_decimated, X_decimated, label=f"Fa' = {normalized_FaD} Hz", color='red')
91 axes[1, 1].axvline(Fa, color='green', linestyle=":", label=f"FaD = {FaD} Hz")
92 axes[1, 1].axvline(normalized_FaD, color='red', label=f"FaD normalized = {normalized_FaD} Hz")
93 axes[1, 1].set_xlabel("Frequency (Hz)")
94 axes[1, 1].set_ylabel("FFT Magnitude")
95 axes[1, 1].set_title("Decimated FFT Spectrum")
96 axes[1, 1].legend()
97 axes[1, 1].grid()
98 |
99 fig.suptitle("Original and Decimated Signals", fontsize=14, fontweight='bold')
100
101 plt.tight_layout()
102 plt.show()

```