

# 6.003: Signal Processing

## Discrete Fourier Transform

analysis

synthesis

**DFT:** 
$$X[k] = \frac{1}{N} \sum_{n=0}^{N-1} x[n] e^{-j\frac{2\pi k}{N}n}$$

$$x[n] = \sum_{k=0}^{N-1} X[k] e^{j\frac{2\pi k}{N}n}$$

**DTFS:** 
$$X[k] = \frac{1}{N} \sum_{n=\langle N \rangle} x[n] e^{-j\frac{2\pi k}{N}n}$$

$$x[n] = \sum_{k=\langle N \rangle} X[k] e^{j\frac{2\pi k}{N}n}$$

**DTFT:** 
$$X(\Omega) = \sum_{n=-\infty}^{\infty} x[n] e^{-j\Omega n}$$

$$x[n] = \frac{1}{2\pi} \int_{2\pi} X(\Omega) e^{j\Omega n} d\Omega$$

## Analyzing Frequency Content of Arbitrary Signals

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Why use a DFT?

- Fourier Series: conceptually simple, but limited to periodic signals.
- Fourier Transforms: arbitrary signals, but continuous domain ( $\Omega$ )
  - good for theory; not so good for numerical evaluation
- Discrete Fourier Transform: arbitrary DT signals, discrete domain ( $k$ )
  - good for computation → broadly used in “Digital Signal Processing”

**Today:** using the DFT to analyze frequency content of a signal.

## Single Sinusoid

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Create four signals

$$x_1[n] = \cos(8\pi n/100)$$

$$x_2[n] = \cos(8\pi n/100 - \pi/4)$$

$$x_3[n] = \cos(9\pi n/100)$$

$$x_4[n] = \cos(9\pi n/100 - \pi/2)$$

each with a duration of 1 second when the sample frequency is 44,100 Hz.

Compare the DFTs of the first 100 samples of each of these signals.

## Python Code

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```
from math import cos, pi
from lib6003.audio import wav_write
from matplotlib.pyplot import stem, show

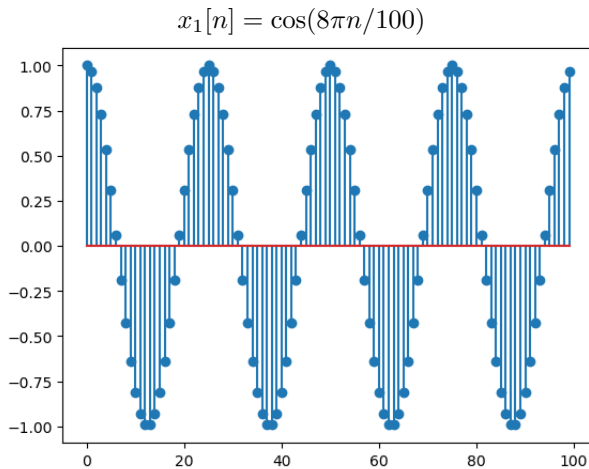
fs = 44100
x1 = [cos(8*pi*n/100) for n in range(fs)]
x2 = [cos(8*pi*n/100-pi/4) for n in range(fs)]
x3 = [cos(9*pi*n/100) for n in range(fs)]
x4 = [cos(9*pi*n/100-pi/2) for n in range(fs)]

wav_write(x1,fs,'x1.wav')
wav_write(x2,fs,'x2.wav')
wav_write(x3,fs,'x3.wav')
wav_write(x4,fs,'x4.wav')

stem(x1[0:100])
show()
```

## Single Sinusoid

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What is the frequency of this tone if the sample rate is 44,100 Hz?

## Single Sinusoid

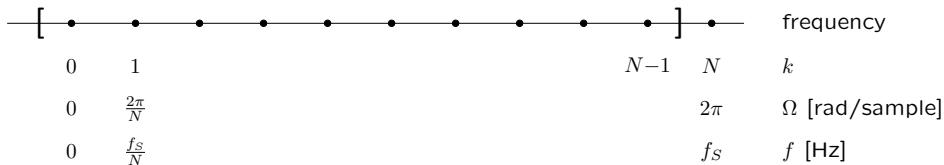
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What is the frequency of the tone generated by  $x_1[n]$ ?

Since  $x_1[n] = \cos(8\pi n/100)$ , we know that the discrete frequency  $\Omega_1 = \frac{8\pi}{100}$ . Furthermore, the sample frequency  $f = f_s$  corresponds to the maximum discrete frequency  $\Omega = 2\pi$ , and frequencies  $f$  in Hz are proportional to discrete frequencies  $\Omega$ .

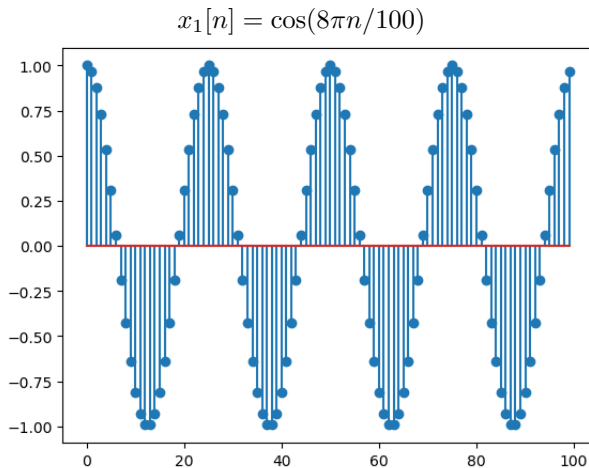
$$\frac{f}{f_s} = \frac{\Omega}{2\pi}$$

$$\text{So } f = \frac{\Omega f_s}{2\pi} = \frac{8\pi/100}{2\pi} \times 44,100 \text{ Hz} = 1764 \text{ Hz}$$



## Single Sinusoid

---



Write a program to calculate the DFT of an input sequence. Use that program to calculate  $X_1[k]$ , which is the DFT of the first 100 samples of  $x_1[n]$ .

## Single Sinusoid

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Use that program to calculate  $X_1[k]$ , which is the DFT of the first 100 samples of  $x_1[n]$ .

```
def dft(x):
    N = len(x)
    answer = [0 for k in range(N)]
    for k in range(N):
        for n in range(N):
            answer[k] += (1/N)*x[n]*e**(-2j*pi*k*n/N)
    return answer

X1 = dft(x1[0:100])
```

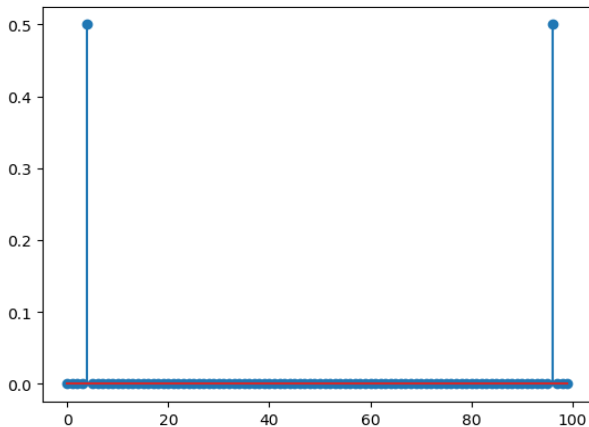


## Single Sinusoid

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Plot the magnitude of  $X_1[\cdot]$ .

$$X_1[k] = DFT\{x_1[0 : 100]\}$$



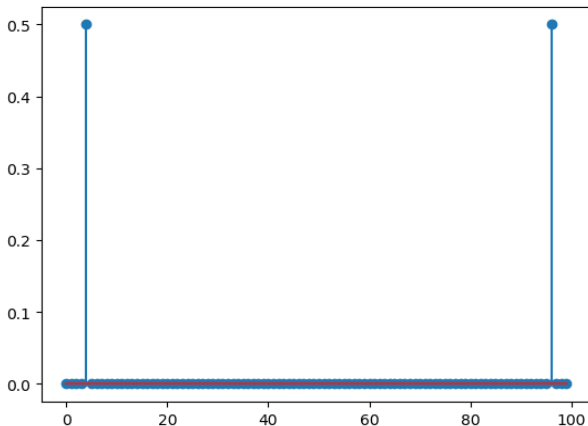
Which values of  $k$  are non-zero?

## Single Sinusoid

---

Plot the magnitude of  $X_1[\cdot]$ .

$$X_1[k] = DFT\{x_1[0 : 100]\}$$



Which values of  $k$  are non-zero?

$$k = \frac{\Omega N}{2\pi} = \frac{8\pi}{100} \times \frac{100}{2\pi} = 4$$

$k = -4$  is also non-zero (Euler's formula).

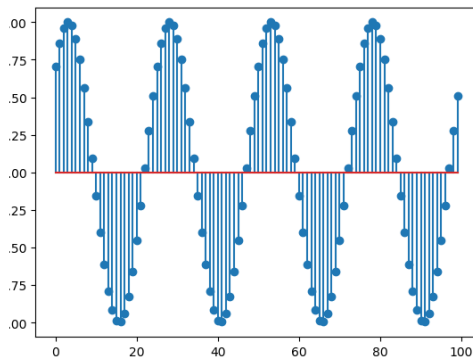
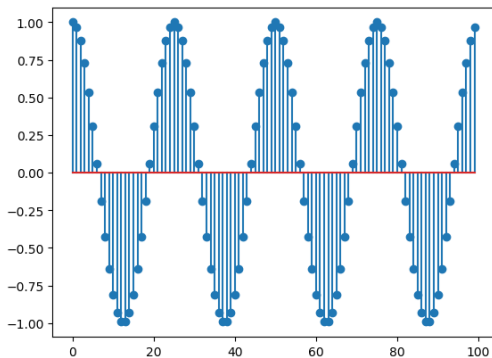
Also  $k = 100 - 4 = 96$  is non-zero since  $X[k]$  is periodic in  $N$ .

## Compare Two Signals

How will plots of DFT magnitudes differ for the following signals?

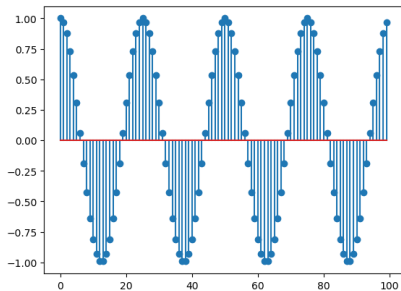
$$x_1[n] = \cos(8\pi n/100)$$

$$x_2[n] = \cos(8\pi n/100 - \pi/4)$$

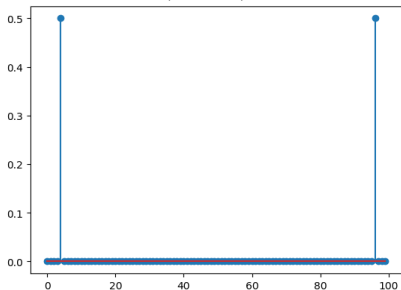


# Compare Two Signals

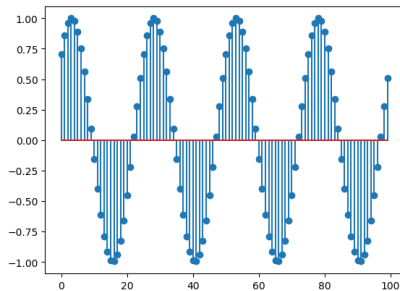
$$x_1[n] = \cos(8\pi n/100)$$



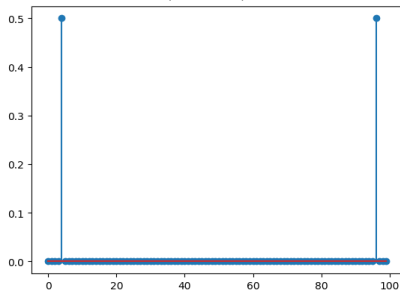
$$|X_1[k]|$$



$$x_2[n] = \cos(8\pi n/100 - \pi/4)$$



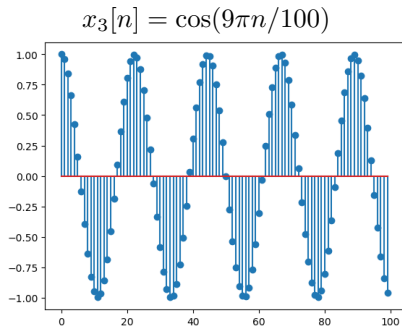
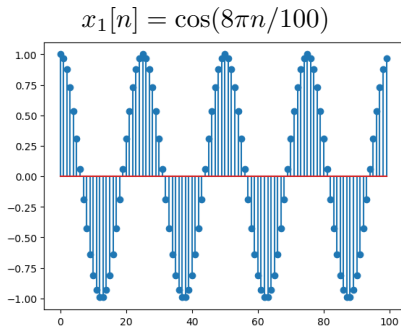
$$|X_2[k]|$$



No difference in magnitudes (but the phases are different).

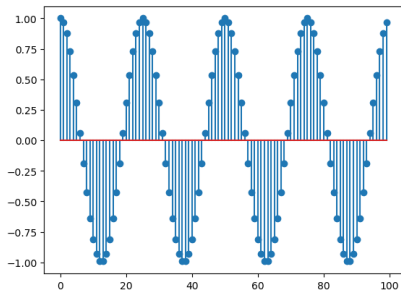
## Compare Two Signals

How will plots of DFT magnitudes differ for the following signals?

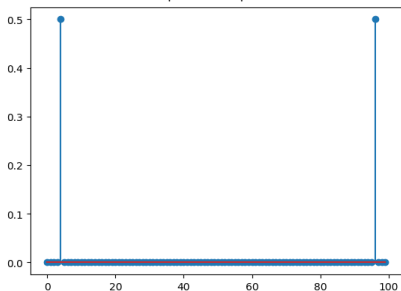


# Compare Two Signals

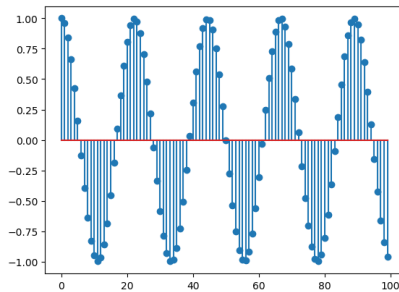
$$x_1[n] = \cos(8\pi n/100)$$



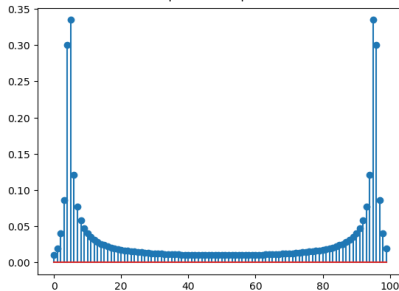
$$|X_1[k]|$$



$$x_3[n] = \cos(9\pi n/100)$$



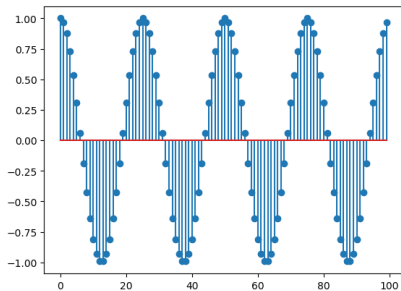
$$|X_3[k]|$$



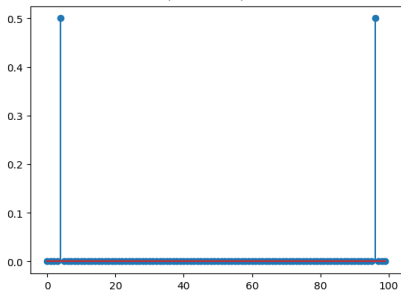
Why are these DFTs so different?

# Compare Two Signals

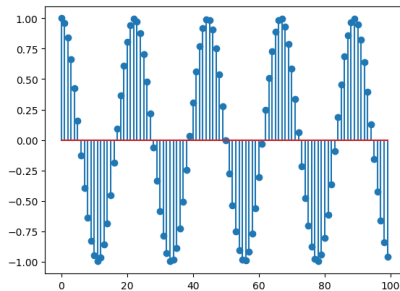
$$x_1[n] = \cos(8\pi n/100)$$



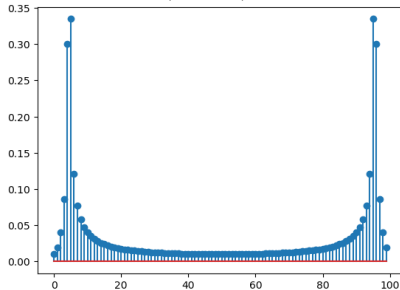
$$|X_1[k]|$$



$$x_3[n] = \cos(9\pi n/100)$$



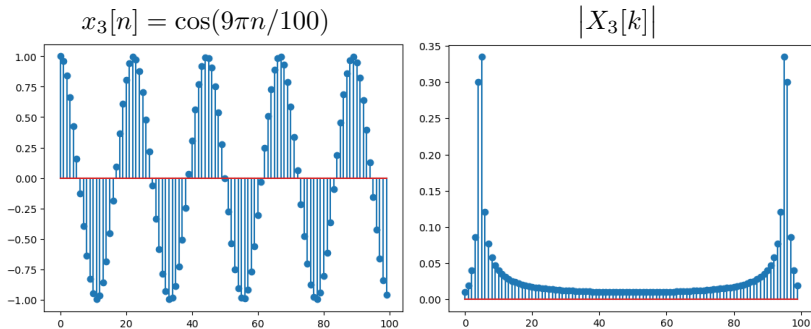
$$|X_3[k]|$$



$\Omega_1 \neq \Omega_3$ . Even more importantly,  $x_3[n]$  is not periodic in  $N = 100$ !

## Single Sinusoid

This blurring occurs because the signal is not periodic in the analysis window ( $N = 100$ ).

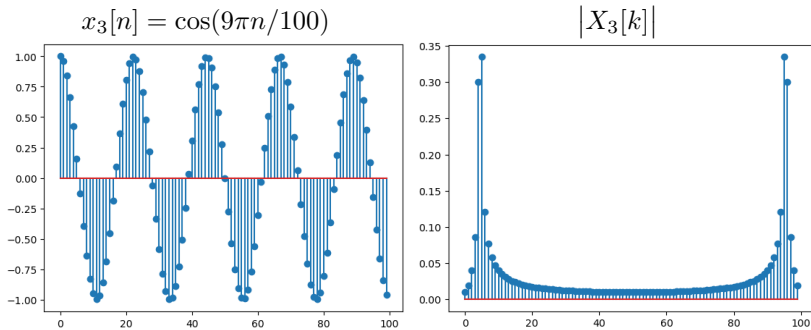


What value of  $k$  corresponds to  $\Omega = 9\pi/100$ ?



## Single Sinusoid

This blurring occurs because the signal is not periodic in the analysis window ( $N = 100$ ).



What value of  $k$  corresponds to  $\Omega = 9\pi/100$ ?

$$\Omega = 9\pi/100 = 2\pi k/N$$

$$k = 4.5$$

The signal frequency fell between the analysis frequencies.

## Compare Two Signals

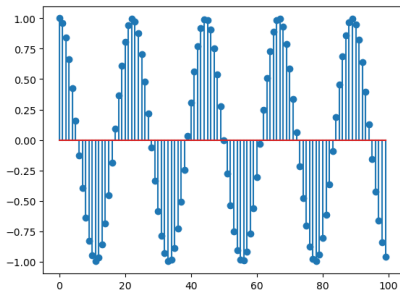
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How will plots of DFT magnitudes differ for the following signals?

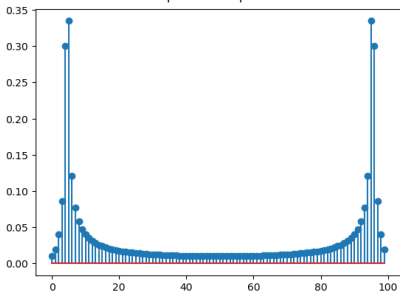
- $x_3[n] = \cos(9\pi n/100)$
- $x_4[n] = \cos(9\pi n/100 - \pi/2)$

# Compare Two Signals

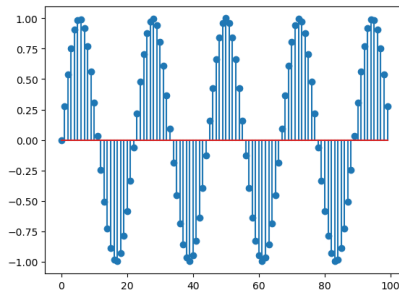
$$x_3[n] = \cos(9\pi n/100)$$



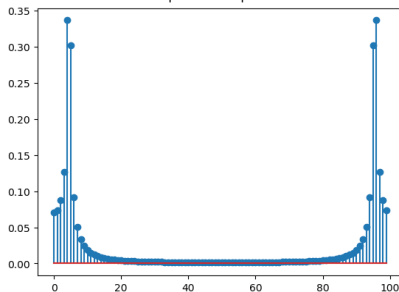
$$|X_3[k]|$$



$$x_4[n] = \cos(9\pi n/100 - \pi/2)$$



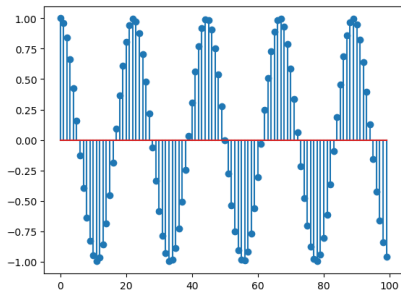
$$|X_4[k]|$$



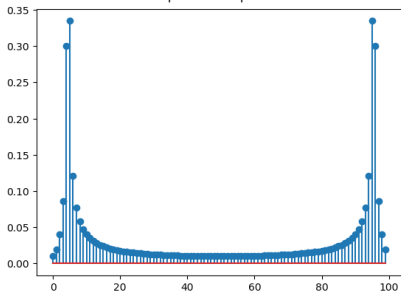
$\Omega_3 = \Omega_4$ . But DC bigger: 5 positive half cycles versus 4 negative ones.

# Compare Two Signals

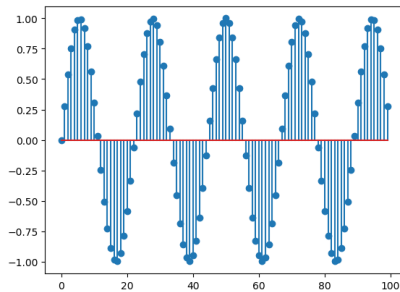
$$x_3[n] = \cos(9\pi n/100)$$



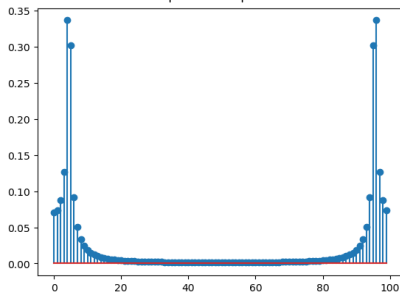
$$|X_3[k]|$$



$$x_4[n] = \cos(9\pi n/100 - \pi/2)$$



$$|X_4[k]|$$



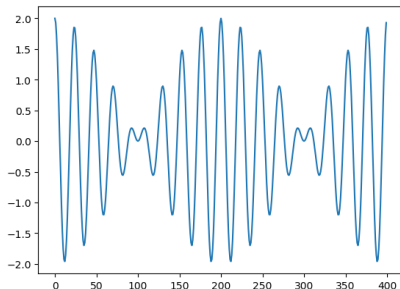
High freq. content of  $X_4$  smaller than  $X_3$ :  $|x_4[99] - x_4[0]| < |x_3[99] - x_3[0]|$

## Analyzing Signals with Multiple Frequencies

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What is the minimum window size  $N$  needed to resolve  $\Omega = 8\pi/100$  from  $9\pi/100$ ?

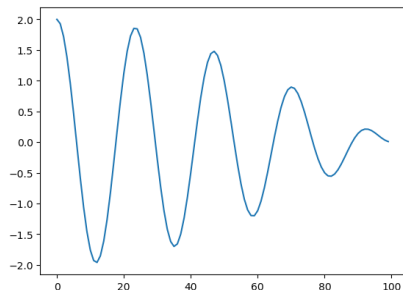
$$x_5[n] = \cos(8\pi n/100) + \cos(9\pi n/100)$$



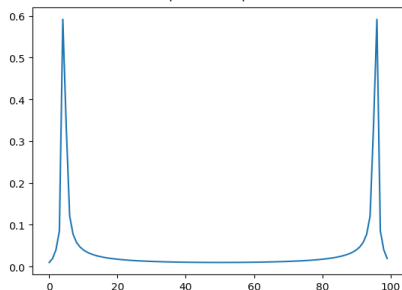
## Analyzing Signals with Multiple Frequencies

If the analysis window is small (here  $N=100$ ), the two frequencies  $8\pi/100$  and  $9\pi/100$  generate a single peak in the DFT at  $k = 4$  (along with its partner at  $k = 100-4 = 96$ ).

$$x_5[n] = \cos(8\pi n/100) + \cos(9\pi n/100)$$



$$|X_5[k]|$$



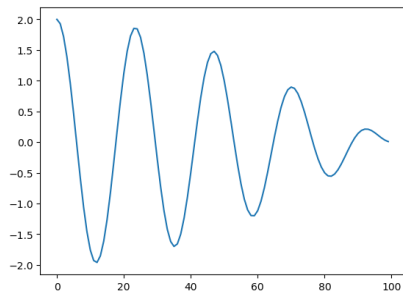
## Analyzing Signals with Multiple Frequencies

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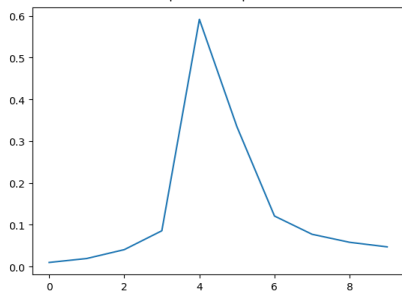
Two frequencies can look like one if analysis window is too small.

$N = 100$  zoomed

$$x_5[n] = \cos(8\pi n/100) + \cos(9\pi n/100)$$



$$|X_5[k]|$$



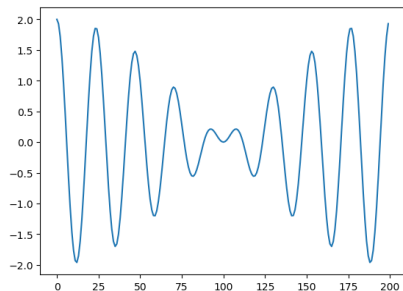
## Analyzing Signals with Multiple Frequencies

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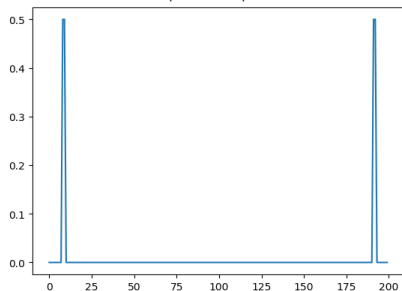
Two frequencies can look like one if analysis window is too small.

$N = 200$

$$x_5[n] = \cos(8\pi n/100) + \cos(9\pi n/100)$$



$$|X_5[k]|$$





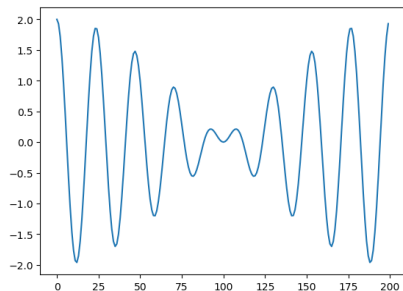
## Analyzing Signals with Multiple Frequencies

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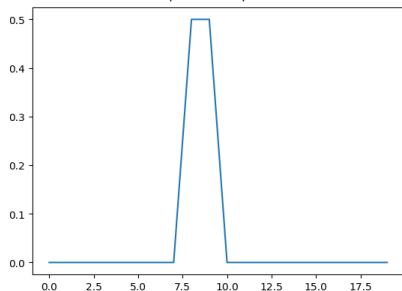
Two frequencies can look like one if analysis window is too small.

$N = 200$  zoomed

$$x_5[n] = \cos(8\pi n/100) + \cos(9\pi n/100)$$



$$|X_5[k]|$$



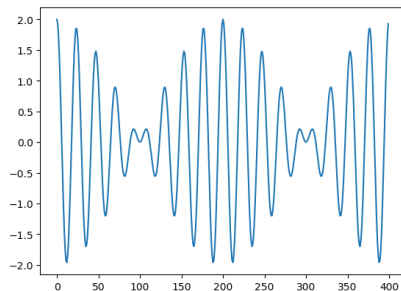
## Analyzing Signals with Multiple Frequencies

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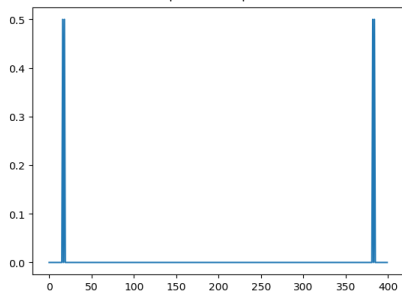
Two frequencies can look like one if analysis window is too small.

$N = 400$

$$x_5[n] = \cos(8\pi n/100) + \cos(9\pi n/100)$$



$$|X_5[k]|$$



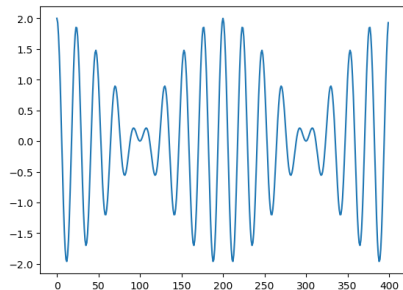
## Analyzing Signals with Multiple Frequencies

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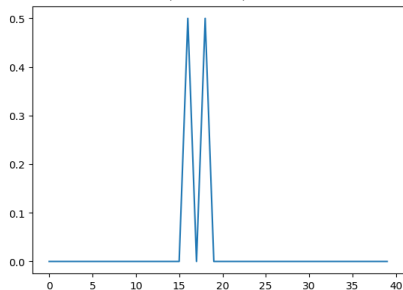
Two frequencies can look like one if analysis window is too small.

$N = 400$  zoomed

$$x_5[n] = \cos(8\pi n/100) + \cos(9\pi n/100)$$



$$|X_5[k]|$$



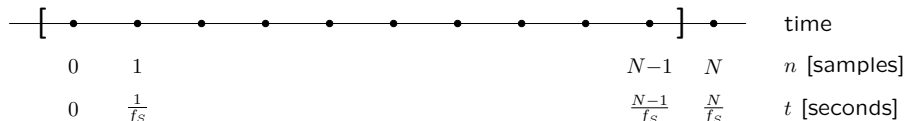
These frequencies are clearly resolved with  $N = 400$ .

## Frequency Scales

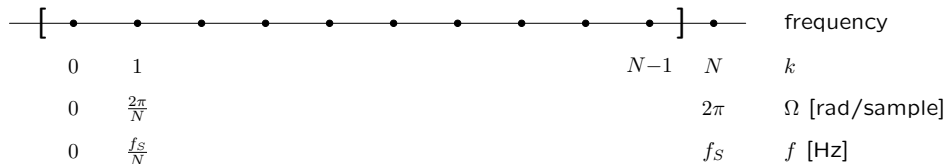
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We can think of the DFT as having spectral resolution of  $(2\pi/N)$  radians, which is equivalent to  $(f_S/N)$  Hz.

The time window is divided into  $N$  samples numbered  $n = 0$  to  $N-1$ .



Discrete frequencies are similarly numbered as  $k = 0$  to  $N-1$ .



## Analyzing Signals with Multiple Frequencies

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Two frequencies are resolved if they are separated by more than  $\frac{2\pi}{N}$ .

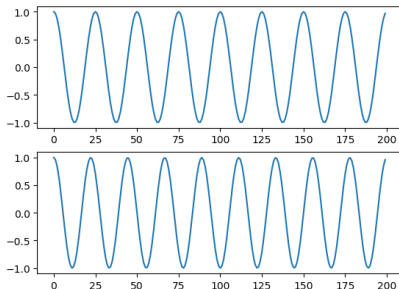
$\Omega_1 = \frac{8\pi}{100}$  and  $\Omega_2 = \frac{9\pi}{100}$  will be resolved if

$$\Delta\Omega = \Omega_2 - \Omega_1 = \frac{9\pi}{100} - \frac{8\pi}{100} = \frac{\pi}{100} > \frac{2\pi}{N}$$

That is, if  $N > 200$ .

We can think of  $\frac{2\pi}{N}$  as the frequency resolution of the DFT.

Notice 8 full cycles of  $\Omega_1$  and 9 full cycles of  $\Omega_2$  fit in  $N = 200$ .



## Summary

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Time and frequency resolution are important issues in all Fourier analyses.

Frequency resolution is determined by the number of samples  $N$  included in the analysis.

