

6.300: Signal Processing

Tutorial Topic

Discrete Fourier Transform

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Discrete Fourier Transform

We may represent a periodic discrete-time signal $x[n] = x[n + N]$ as a finite-length sum of harmonically-related complex exponentials with a discrete-time Fourier series. Periodicity is rather restrictive, as many real-world signals of interest are not periodic. To be periodic, a signal must repeat forever!

The discrete-time Fourier transform is a Fourier representation for aperiodic discrete-time signals $x[n]$, but the analysis formula entails an infinite-length sum, and the spectrum $X(\Omega)$ — a function of the continuous variable Ω — is continuous. While a useful theoretical tool, the DTFT is also somewhat impractical.

In search of a Fourier representation that is amenable for digital computation, we turn to the DFT. You might call the DFT a discrete-time, discrete-frequency Fourier transform.

- finite-length signals
- discrete in time
- discrete in frequency

windowing $x_w[n] = x[n]w[n]$
time indexed by integer n
frequency indexed by integer k

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

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

Note: There is no general consensus on where to place the $1/N$ factor in the equations that define the DFT. In 6.300, we place the $1/N$ factor in the analysis equation, so that the DFT and DTFS analysis equations match. In both cases, then, the DC term $X[0]$ is the average value of $x[n]$ over a length- N interval. Some authors and numerical packages (e.g., MATLAB) place the $1/N$ factor in the synthesis equation instead, however. Other authors and numerical packages even put a factor of $1/\sqrt{N}$ in both the analysis and synthesis equations to make the DFT a unitary transformation.

Relation to Discrete-Time Fourier Series (DTFS)

The length- N DFT is equivalent to the DTFS of an N -periodic extension of the windowed signal $x_w[n] = x[n]w[n]$. From this perspective, sharp discontinuities in the periodic extension of $x_w[n]$ lead to spurious high-frequency content spread across the spectrum of $x_w[n]$.

$$X[k] = \frac{1}{N} \sum_{n=0}^{N-1} x_w[(n \bmod N)] e^{-j2\pi kn/N}$$

Relation to Discrete-Time Fourier Transform (DTFT)

The DFT returns N scaled, equally-spaced samples of the DTFT over the interval $[0, 2\pi)$. Increasing the DFT length N yields more samples of the DTFT, which are spaced more closely together.

$$X[k] = \frac{1}{N} X_w\left(\frac{2\pi}{N} k\right)$$

cyclical frequency resolution: $\frac{f_s}{N}$ hertz

angular frequency resolution: $\frac{2\pi}{N}$ radians

Single Sinusoid

Create four signals

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

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

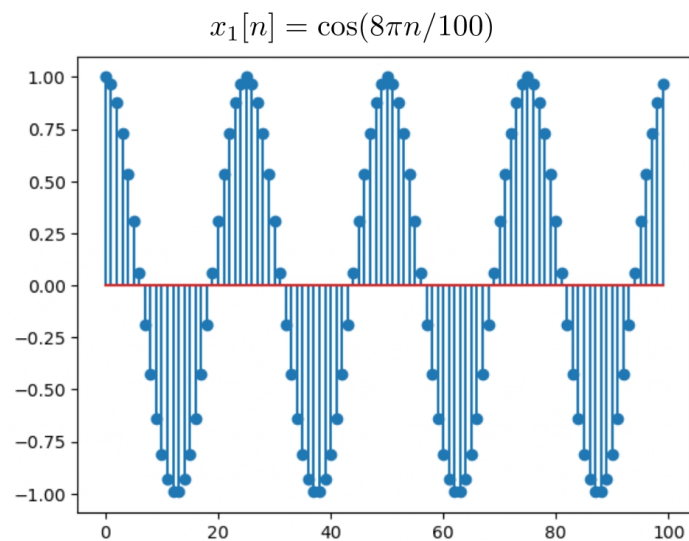
$$x_3[n] = \cos(9\pi n/100)$$

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

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.

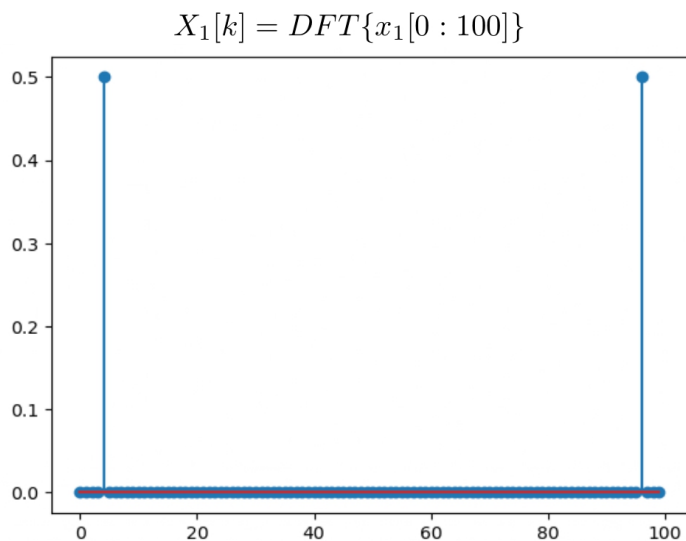
Single Sinusoid



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

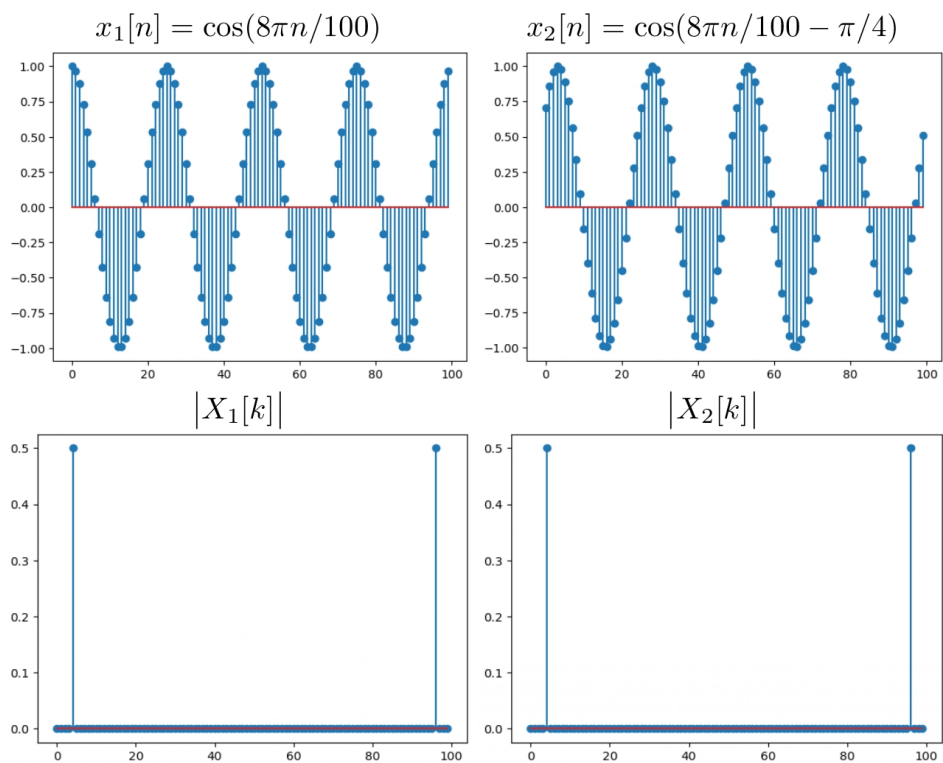
Single Sinusoid

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



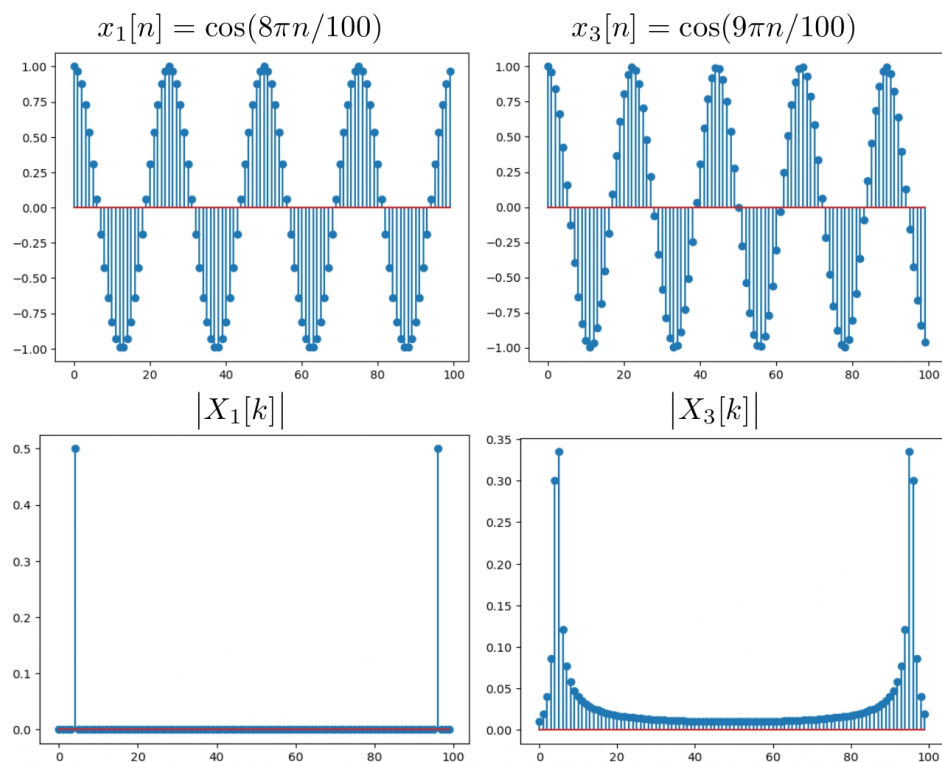
Which values of k are non-zero?

Compare Two Signals



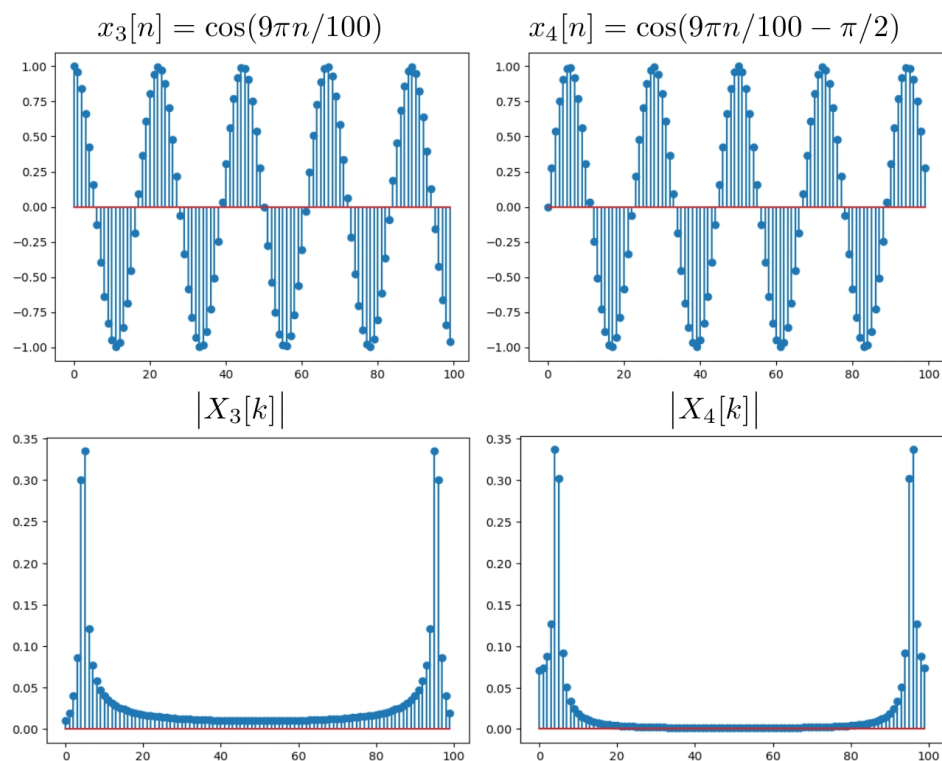
No difference in magnitudes (but the phases are different).

Compare Two Signals



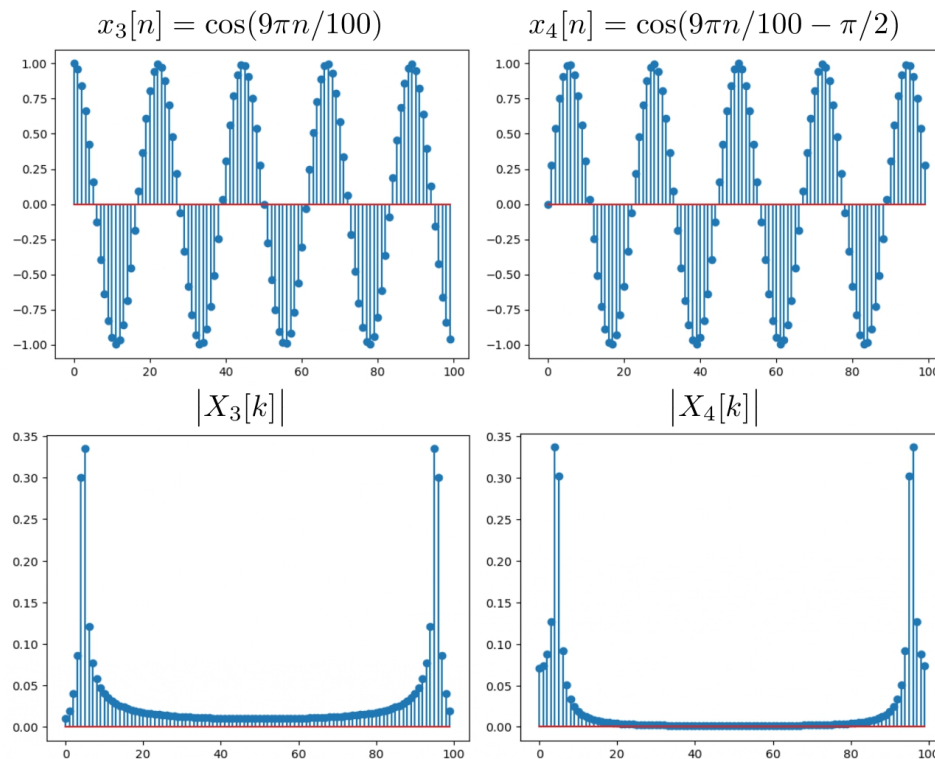
Why are these DFTs so different?

Compare Two Signals



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

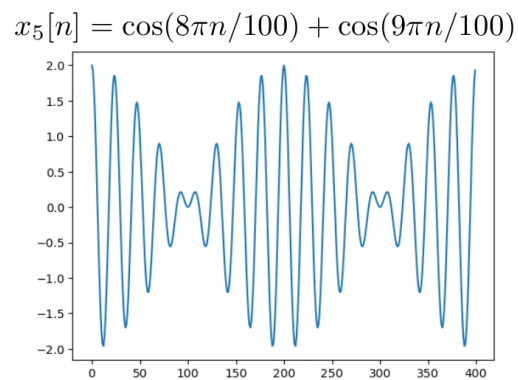
Compare Two Signals



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

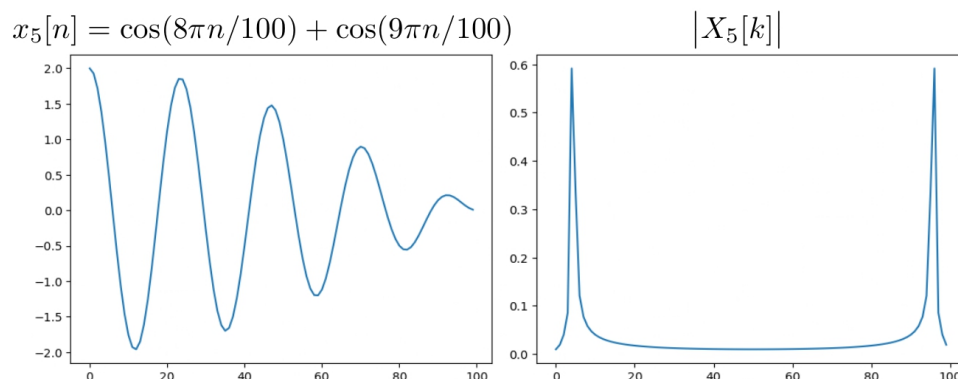
Analyzing Signals with Multiple Frequencies

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



Analyzing Signals with Multiple Frequencies

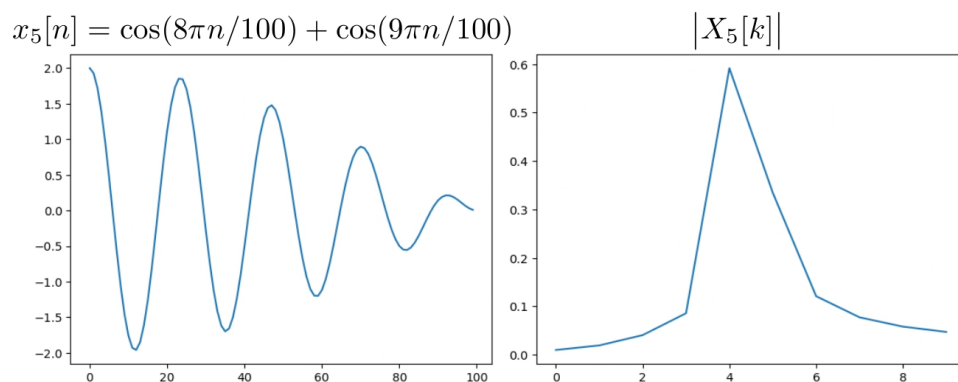
If the analysis window is small (here $N=100$), the two frequencies $8\pi/100$ and $9\pi/100$ generate a single peak in the DFT at $k = 4$ (along with its partner at $k = 100-4 = 96$).



Analyzing Signals with Multiple Frequencies

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

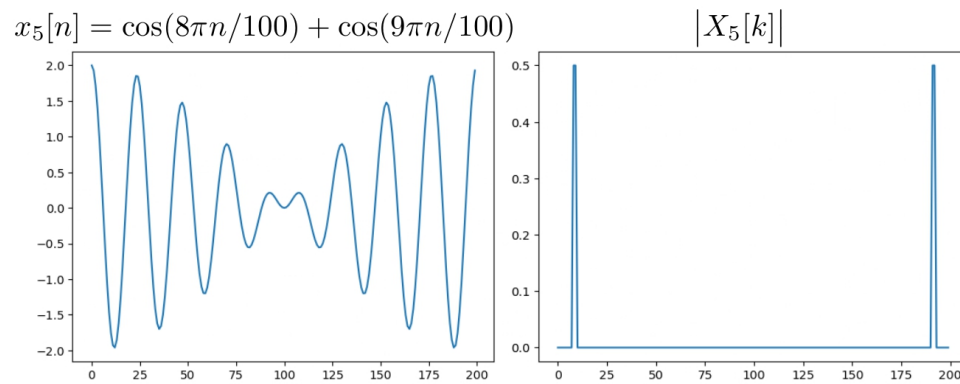
$N = 100$ zoomed



Analyzing Signals with Multiple Frequencies

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

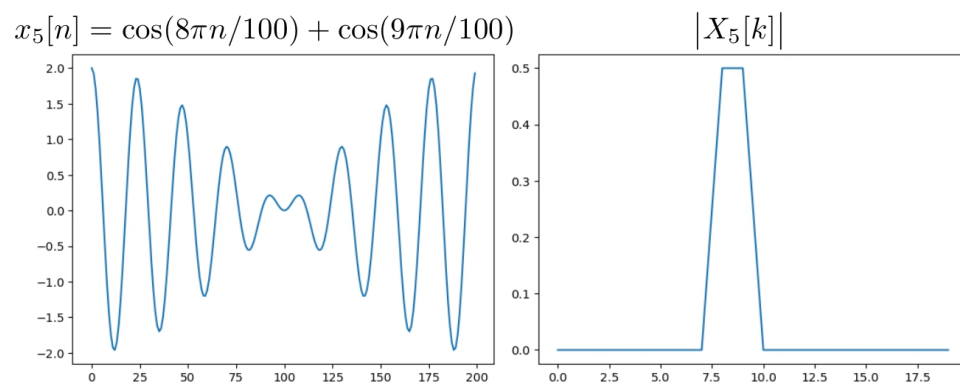
$N = 200$



Analyzing Signals with Multiple Frequencies

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

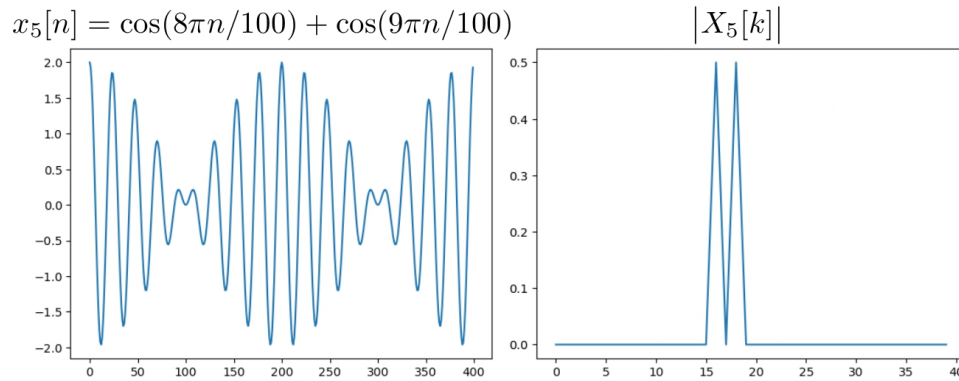
$N = 200$ zoomed



Analyzing Signals with Multiple Frequencies

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

$N = 400$ zoomed

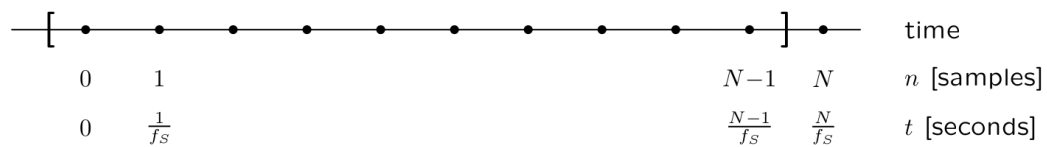


These frequencies are clearly resolved with $N = 400$.

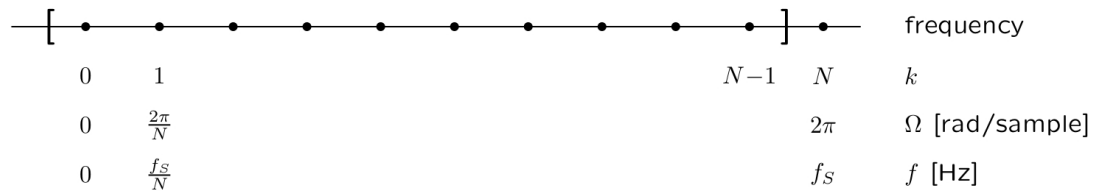
Frequency Scales

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

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

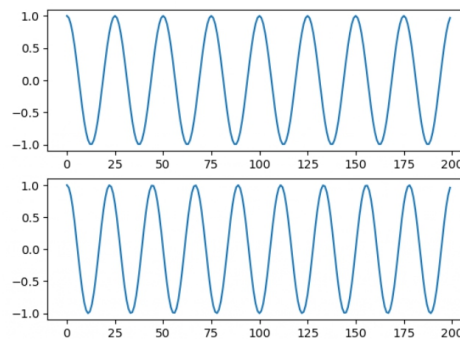
$\Omega_1 = \frac{8\pi}{100}$ and $\Omega_2 = \frac{9\pi}{100}$ will be resolved if

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

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 Ω_1 and 9 full cycles of Ω_2 fit in $N = 200$.



Summary

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.

