# 6.3000: Signal Processing

### **Discrete Fourier Transform**

#### analysis

#### synthesis

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

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

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

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

$$\label{eq:dispersive} \mathbf{DTFT:} \qquad X(\Omega) = \sum_{n=-\infty}^{\infty} x[n] e^{-j\Omega n} \qquad \qquad x[n] = \frac{1}{2\pi} \int_{2\pi} X(\Omega) e^{j\Omega n} d\Omega$$

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### **Analyzing Frequency Content of Arbitrary Signals**

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.

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  $44,100\,\mathrm{Hz}.$ 

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

### **Python Code**

from math import cos, pi
from lib6003.audio import wav\_write
from matplotlib.pyplot import stem, show

fs = 44100

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

```
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()
```



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

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

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  $\Omega$ .





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  $x_1[n]$ .

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```
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])

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



Which values of k are non-zero?

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



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

How will plots of DFT magnitudes differ for the following signals?  $x_1[n] = \cos(8\pi n/100)$   $x_2[n] = \cos(8\pi n/100 - \pi/4)$ 





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

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





Why are these DFTs so different?



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



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

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.

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

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



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



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

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



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).



Two frequencies can look like one if analysis window is too small.  ${\cal N}=100 \ {\rm zoomed}$ 



Two frequencies can look like one if analysis window is too small.  ${\cal N}=200$ 



Two frequencies can look like one if analysis window is too small.  ${\cal N}=200~{\rm zoomed}$ 



Two frequencies can look like one if analysis window is too small.  ${\cal N}=400$ 



Two frequencies can look like one if analysis window is too small.  ${\cal N}=400~{\rm zoomed}$ 



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  $(f_S/N)\,{\rm Hz}.$ 

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.



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

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

Frequency resolution is determined by the number of samples  ${\cal N}$  included in the analysis.

