

# **ELEC 221 Lecture 09**

## **DT Fourier series; filters**

Thursday 03 October 2024

## Announcements

Q2.5 is updated;  
↗ please check PrairieLearn

- Assignment 2 due Saturday 23:59 (final question removed, deferred to A3)
- Midterm 1 on Tuesday (bring your student ID and writing implements)


## Last time

We explored periodic DT complex exponential signals:

$$x[n] = e^{j\omega n} = e^{j \cdot \frac{2\pi}{N} n} \quad \hookrightarrow \text{fundamental period}$$

We found that these signals behave differently than CT signals...

**Difference 1:** we only need to consider  $\omega$  in the range  $[0, 2\pi)$ .

$$\begin{aligned} x[n] &= e^{j(\omega + 2\pi)n} \\ &= e^{j\omega n} \cdot \underbrace{e^{j2\pi n}}_1 \\ &= e^{j\omega n} \end{aligned}$$


**Difference 2:** there are additional criteria for periodicity.

$$\begin{aligned}
 x[n+N] &= e^{j\omega(n+N)} \\
 &= e^{j\omega n} \cdot \underbrace{e^{j\omega N}}_1
 \end{aligned}
 \quad \omega N = 2\pi \cdot m$$

$\downarrow$   
 $m$  is integer

Example:  $x[n] = \sin(5\pi n/7)$  is periodic.

- In CT, period of  $x(t) = \sin(5\pi t/7)$  is  $T = \frac{2\pi}{\omega} = \frac{14}{5}$
- In DT, period of  $x[n] = \sin(5\pi n/7)$  is  $N=14$

Example:  $x[n] = \sin(5n/7)$  is NOT periodic in DT.

$$\frac{5\pi n}{7} = 2\pi \cdot m$$

$$\sum_{k=-\infty}^{\infty} e^{j\omega_k t}$$

**Difference 3:** there are only finitely many harmonics.

$$x_0[n] = 1$$

$$x_1[n] = e^{j \cdot \frac{2\pi}{N} n}$$

$$x_2[n] = e^{j \cdot 2 \cdot \frac{2\pi}{N} n}$$

$$\vdots$$

$$x_{N-1}[n] = e^{j(N-1) \cdot \frac{2\pi}{N} n}$$

$$x_N[n] = e^{jN \frac{2\pi}{N} n} = e^{j \cdot 2\pi n} = 1$$

$$\vdots$$

$$\Rightarrow x_k[n] = e^{j \frac{2\pi}{N} k n}$$

## Last time

We found DT complex exponential signals are also eigenfunctions of LTI systems.

$$\begin{aligned}x(t) &= e^{j\omega t} \rightarrow y(t) = H(j\omega) \cdot e^{j\omega t} \\y[n] &= \sum_{k=-\infty}^{\infty} x[n-k] h[k] & x[n] &= e^{jm\omega n} \\&= \sum_{k=-\infty}^{\infty} e^{jm\omega(n-k)} h[k] \\&= e^{jm\omega n} \sum_{k=-\infty}^{\infty} e^{-jkm\omega} h[k] \\&= x[n] \cdot H(e^{j\omega})\end{aligned}$$

We need a Fourier series representation of DT signals:

$$x[n] = \sum_{k=0}^{N-1} c_k e^{jk\omega n}$$

## Learning outcomes:

- Evaluate Fourier series coefficients of DT signals
- Define filter systems through the frequency response
- ~~Distinguish between finite impulse response and infinite impulse response filters in DT~~

# DT Fourier coefficients

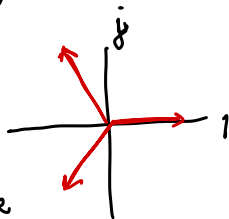
$$x[n] = \sum_{k=0}^{N-1} c_k e^{jk\omega n}$$

$$x(t) = \sum_{k=-\infty}^{\infty} c_k e^{jk\omega t}$$

$$1 + e^{j\frac{2\pi}{5}n} + e^{j2 \cdot \frac{2\pi}{5}n} + \dots$$

Leverage the following identity about complex numbers:

$$\sum_{n=0}^{N-1} e^{jk\frac{2\pi}{N}n} = \begin{cases} N & \text{if } k=0, \pm N, \pm 2N, \dots \\ 0 & \text{otherwise} \end{cases}$$



We will multiply on both sides, and sum.

$$N=2: \quad e^{j\pi \cdot 0} = 1 \quad e^{j\pi \cdot 1} = -1 \quad e^{j\frac{2\pi}{N}k}$$

$$N=3: \quad e^{j\frac{2\pi}{3} \cdot 0} \quad e^{j\frac{2\pi}{3}} \quad e^{j\frac{4\pi}{3}} \quad e^{j\frac{2\pi}{N}k}$$



# DT Fourier coefficients

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

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

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

$$= \sum_{n=0}^{N-1} \sum_{k=0}^{N-1} C_k e^{jn \frac{2\pi}{N} (k-m)}$$

$$= C_m \cdot N$$

$$C_m = \frac{1}{N} \sum_{n=0}^{N-1} x[n] e^{-jm \frac{2\pi}{N} n}$$

$$x(t) = \sum C_k e^{jk\omega t}$$

$$C_k = \frac{1}{T} \int_T x(t) e^{-jk\omega t} dt$$



# DT Fourier coefficients

DT Fourier synthesis equation

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

DT Fourier analysis equation

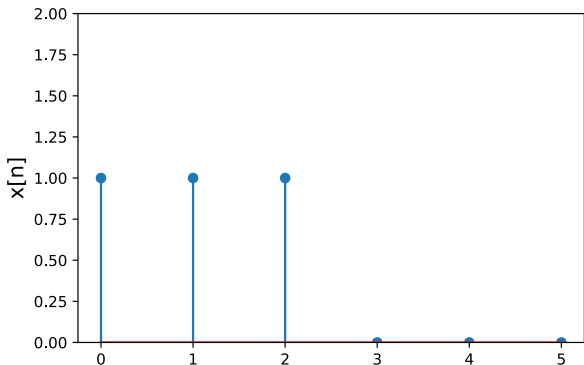
$$c_k = \frac{1}{N} \sum_{n=0}^{N-1} x[n] e^{-jk \frac{2\pi}{N} n} = \frac{1}{N} \sum_{n=\langle N \rangle} x[n] e^{-jk \frac{2\pi}{N} n}$$

$N=5$

|          |          |          |       |       |       |         |         |       |       |       |         |
|----------|----------|----------|-------|-------|-------|---------|---------|-------|-------|-------|---------|
| $c_{-3}$ | $c_{-2}$ | $c_{-1}$ | $c_0$ | $c_1$ | $c_2$ | $c_3$   | $c_4$   | $c_5$ | $c_6$ | $c_7$ | $\dots$ |
|          | $c_2^*$  | $c_1^*$  |       |       |       | $c_2$   | $c_1$   | $c_0$ | $c_1$ | $c_2$ | $\dots$ |
|          |          |          |       |       |       | $c_2^*$ | $c_1^*$ |       |       |       |         |

## Exercise: the DT square wave

Compute the Fourier coefficients of this signal:



$$C_k = \frac{1}{6} \sum_{n=0}^5 x[n] e^{-jk\frac{2\pi}{6}n} = \frac{1}{6} \sum_{n=0}^5 x[n] e^{-jk\frac{\pi}{3}n}$$

$N=6$

Exercise: the DT square wave

$$C_k = \frac{1}{6} \sum_{n=0}^5 x[n] e^{-j \frac{k\pi n}{3}}$$

$$C_0 = \frac{1}{6} \sum_{n=0}^5 x[n] = \frac{1}{6} [1+1+1+0+0+0] = \frac{1}{2}$$

$$C_1 = \frac{1}{6} \sum_{n=0}^5 x[n] e^{j \frac{\pi n}{3}} = \frac{1}{6} [1 + e^{j \frac{\pi}{3}} + e^{j \frac{2\pi}{3}}] = \frac{1}{6} [1 - \sqrt{3}j] \Rightarrow C_5$$

$$C_2 = \frac{1}{6} \sum_{n=0}^5 x[n] e^{j \frac{2\pi n}{3}} = \frac{1}{6} [1 + e^{j \frac{2\pi}{3}} + e^{j \frac{4\pi}{3}}] = 0$$

$$C_3 = \frac{1}{6} \sum_{n=0}^5 x[n] e^{j \pi n} = \frac{1}{6} [1 + e^{j \pi} + e^{j 2\pi}] = \frac{1}{6}$$

$$C_4 = 0$$

$$C_0 \quad C_1 \quad C_2 \quad C_3 \quad C_2^* \quad C_1^*$$

$$C_5 = \frac{1}{6} [1 + \sqrt{3}j] = C_1^*$$

# Properties of DT Fourier coefficients

**TABLE 3.2** PROPERTIES OF DISCRETE-TIME FOURIER SERIES

| Property                               | Periodic Signal  | Fourier Series Coefficients  |
|--|--|--|
|  | $\left. \begin{array}{l} x[n] \\ y[n] \end{array} \right\} \begin{array}{l} \text{Periodic with period } N \text{ and} \\ \text{fundamental frequency } \omega_0 = 2\pi/N \end{array}$ | $\left. \begin{array}{l} a_k \\ b_k \end{array} \right\} \begin{array}{l} \text{Periodic with} \\ \text{period } N \end{array}$                            |
| Linearity                              | $Ax[n] + By[n]$  | $Aa_k + Bb_k$  |
| Time Shifting                          | $x[n - n_0]$   | $a_k e^{-jk(2\pi/N)n_0}$   |
| Frequency Shifting                     | $e^{jM(2\pi/N)n} x[n]$   | $a_{k-M}$  |
| Conjugation                            | $x^*[n]$   | $a_{-k}^*$   |
| Time Reversal                          | $x[-n]$  | $a_{-k}$   |
| Time Scaling                           | $x_{(m)}[n] = \begin{cases} x[n/m], & \text{if } n \text{ is a multiple of } m \\ 0, & \text{if } n \text{ is not a multiple of } m \end{cases}$ (periodic with period $mN$ )          | $\frac{1}{m} a_k \left( \begin{array}{l} \text{viewed as periodic} \\ \text{with period } mN \end{array} \right)$  |
| Periodic Convolution                   | $\sum_{r=(N)} x[r]y[n-r]$  | $Na_k b_k$   |
| Multiplication                         | $x[n]y[n]$   | $\sum_{l=(N)} a_l b_{k-l}$   |
| First Difference                       | $x[n] - x[n-1]$  | $(1 - e^{-jk(2\pi/N)})a_k$   |
| Running Sum                            | $\sum_{k=-\infty}^n x[k] \left( \begin{array}{l} \text{finite valued and periodic only} \\ \text{if } a_0 = 0 \end{array} \right)$   | $\left( \frac{1}{1 - e^{-jk(2\pi/N)}} \right) a_k$   |
| Conjugate Symmetry for Real Signals    | $x[n]$ real  | $\begin{cases} a_k = a_{-k}^* \\ \Re\{a_k\} = \Re\{a_{-k}\} \\ \Im\{a_k\} = -\Im\{a_{-k}\} \\  a_k  =  a_{-k}  \\ \angle a_k = -\angle a_{-k} \end{cases}$ |
| Real and Even Signals                  | $x[n]$ real and even   | $a_k$ real and even  |
| Real and Odd Signals                   | $x[n]$ real and odd  | $a_k$ purely imaginary and odd   |
| Even-Odd Decomposition of Real Signals | $\begin{cases} x_e[n] = \mathcal{E}\{x[n]\} & [x[n] \text{ real}] \\ x_o[n] = \mathcal{O}\{x[n]\} & [x[n] \text{ real}] \end{cases}$   | $\begin{cases} \Re\{a_k\} \\ j\Im\{a_k\} \end{cases}$  |

## Exercise: the DT square wave

Let's try the same thing as we did in CT:

- shift the signal left by 1
- speed it up by 2

*try it yourself!*

## Where do we go from here?

We've showed a couple important things so far.

Signals can be expressed in terms of weighted, shifted impulses.

$$x(t) = \int_{-\infty}^{\infty} x(\tau) \delta(t-\tau) d\tau \quad x[n] = \sum_{k=-\infty}^{\infty} x[k] \delta[n-k]$$

## Where do we go from here?

If we know what an LTI system does to a unit impulse (the impulse response  $h(t)$  or  $h[n]$ ), we can learn what it does to any signal.

This was the convolution integral and sum:

$$y(t) = \int_{-\infty}^{\infty} x(\tau) h(t-\tau) d\tau$$
$$y[n] = \sum_{k=-\infty}^{\infty} x[k] h[n-k]$$



## Where do we go from here?

Complex exponential signals are eigenfunctions of LTI systems:

$$x(t) = e^{j\omega t} \rightarrow y(t) = H(j\omega) e^{j\omega t} \quad H(j\omega) = \int_{-\infty}^{\infty} h(\tau) e^{-j\omega\tau} d\tau$$

$$x[n] = e^{j\omega n} \rightarrow y[n] = H(e^{j\omega}) e^{j\omega n} \quad H(e^{j\omega}) = \sum_{k=-\infty}^{\infty} h[k] e^{-j\omega k}$$

$H(j\omega)$  in CT, and  $H(e^{j\omega})$  in DT, are the **frequency response** of the system (more generally, system functions).

$$x(t) = \sum_{k=-\infty}^{\infty} C_k e^{j\omega_k t} \longrightarrow y(t) = \sum_{k=-\infty}^{\infty} C_k H(j\omega_k) e^{j\omega_k t}$$

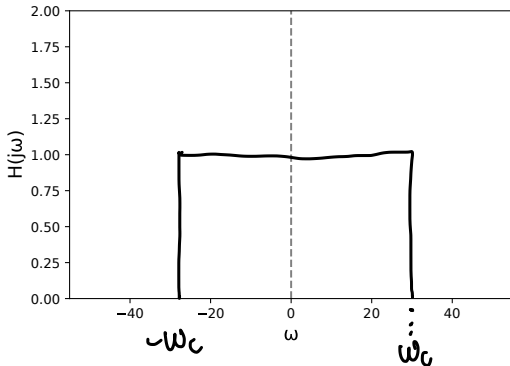
Through careful choice of  $H(j\omega)$  or  $H(e^{j\omega})$ , we can change the behaviour of a system.

## Example

What does a system with the following frequency response do?

$$H(j\omega) = \begin{cases} 1 & |\omega| \leq \omega_c \\ 0 & |\omega| > \omega_c \end{cases}$$

$$y(t) = \sum_{k=-\infty}^{\infty} C_k H(jk\omega) e^{jk\omega t}$$



Filters are LTI systems that can be used to separate out, combine, or modify the components of a signal at specific frequencies.

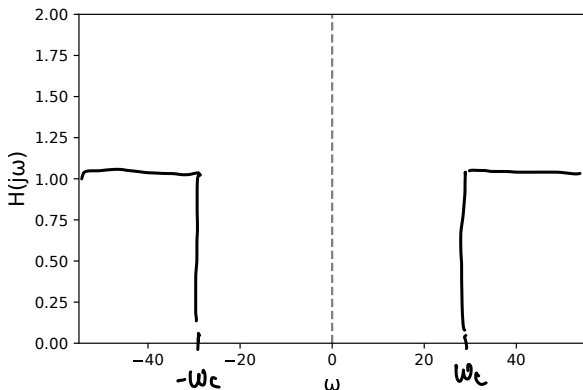
Two key types:

- **Frequency-shaping**: change the amplitudes of parts of a signal at specified frequencies
- **Frequency-selective**: eliminate or attenuate parts of a signal at specified frequencies

## CT frequency-selective filters

We can also consider an ideal **highpass filter**:

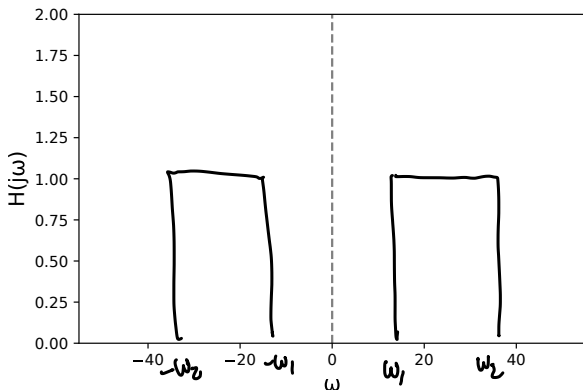
$$H(j\omega) = \begin{cases} 1 & |\omega| > \omega_c \\ 0 & |\omega| \leq \omega_c \end{cases}$$



## CT frequency-selective filters

Or an ideal **bandpass** filter:

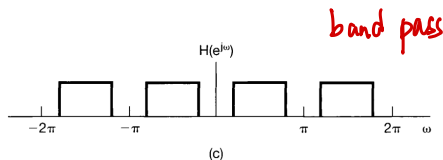
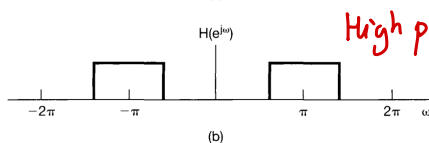
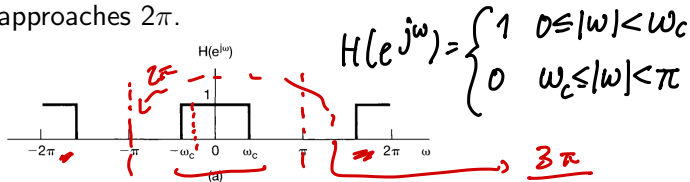
$$H(j\omega) = \begin{cases} 1 & \omega_1 \leq |\omega| \leq \omega_2 \\ 0 & \text{otherwise} \end{cases}$$



# DT filters

Recall that in DT, the frequency increases up until  $\omega = \pi$ , then decreases as it approaches  $2\pi$ .

low pass



$$\frac{3\pi}{2}$$
$$\downarrow$$
$$-\frac{\pi}{2}$$

Today's learning outcomes were:

- Evaluate Fourier series coefficients of DT signals
- Define filter systems through the frequency response
- ~~Distinguish between finite impulse response and infinite impulse response filters in DT~~

## For next time

### Action items:

1. Assignment 2 due Saturday 23:59
2. Study for Midterm 1
3. Suggest tutorial topics on Piazza

### Recommended reading:

- From today's class: Oppenheim 3.6-3.12
- Suggested problems: 3.2, 3.10-3.17, 3.27-3.31, 3.39