

ELEC 221 Lecture 05

The discrete-time Fourier series

Thursday 22 September 2022

- Assignment 2 available; due next Wednesday

Last time

We introduced the Dirichlet conditions, which are sufficient conditions for a CT signal to be expanded as a Fourier series.

If over one period, a signal

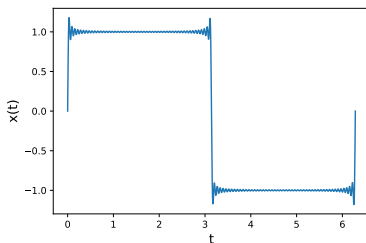
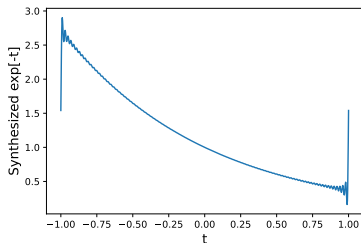
1. is single-valued
2. is absolutely integrable
3. has a finite number of maxima and minima
4. has a finite number of discontinuities

then the Fourier series converges to

- $x(t)$ where it is continuous
- half the value of the jump where it is discontinuous

Last time

We encountered the Gibbs phenomenon, a ringing that occurs when you truncate the Fourier series representation.



We derived that the amount of overshoot was about 9% the height of the discontinuity, no matter how large you make N .

We are shifting over to **discrete time**.

Learning outcomes:

- Compute the fundamental period and frequency of a DT signal
- Express DT periodic signals as Fourier series and compute the Fourier coefficients
- Compute the frequency response of a system, and determine how it responds to a complex exponential signal

DT complex exponential signals

Recall our continuous-time representation of complex exponential signals:

where α could be real or complex.

In DT, we instead often write

where β can be real or complex.

DT complex exponential signals

Case: $x[n] = C\beta^n$, where β is real.

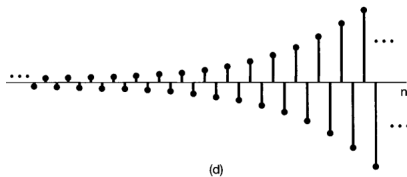
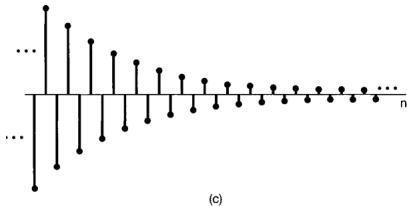
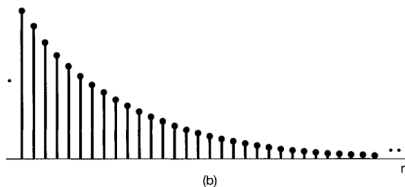
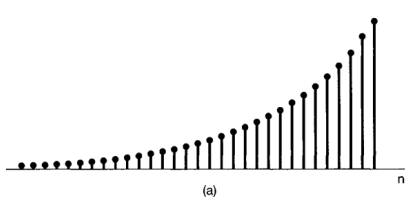
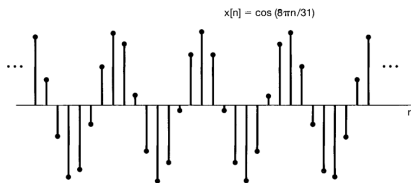
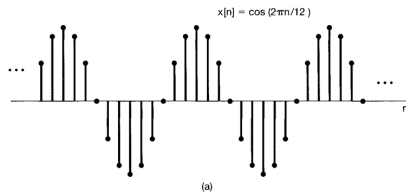


Image credit: Oppenheim chapter 1.3.

DT complex exponential signals

Case: $x[n] = C\beta^n$, where β is purely complex:



Frequency and period of DT complex exponential signals

While these might look similar to their discrete counterparts, there is a **very important difference** relating to frequency.

In CT,

This has frequency ω is periodic with period $T = 2\pi/\omega$.

The bigger ω gets, the faster it oscillates!

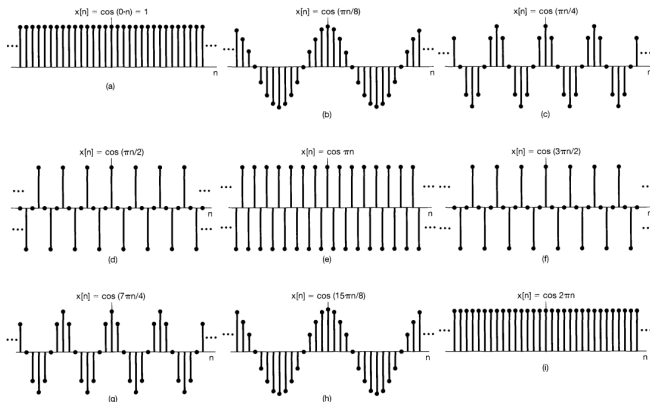
Frequency and period of DT complex exponential signals

What happens in DT? Imagine we have some starting ω :

Suppose that we slowly increase ω . Eventually we reach $\omega + 2\pi$:

Frequency and period of DT complex exponential signals

For a DT signal with frequency ω , the signals with frequencies *are the same!*



Only need to consider ω in the range 0 to 2π (or $-\pi$ to π).

Frequency and period of DT complex exponential signals

Another consequence: there are additional criteria for periodicity.

Suppose the period is N :

This implies that

must be rational for the signal to be periodic.

Frequency and period of DT complex exponential signals

This can be counterintuitive.

Example: consider the CT signal

This is periodic with period $T = 2\pi/3$.

What about the DT equivalent,

Frequency and period of DT complex exponential signals

The signal

is *not periodic*.

n can only be an integer, so we will never get to the “time” of $2\pi/3$ where the signal will start to repeat.

Example: what is the fundamental period of

For the CT equivalent, $T = 2\pi/\omega = 12/5$, but this is not a valid n .

But after 5 periods, we are at $n = 12$ which is valid so the fundamental period is $N = 12$.

What about harmonics?

In CT we had an infinite number of these. What about DT?

Note that, once we reach $k = N$, we get back to our original signal... so have only N distinct harmonics:

DT signals and LTI systems

Recall that in CT, complex exponentials are eigenfunctions of LTI systems. What about in DT?

Consider a system with impulse response $h[n]$ and DT signal $x_m[n] = e^{jm\omega n}$. Use the convolution sum:

So in DT, if we know how the system responds to complex exponential signals, we can figure out what it does to any signal that is expressed in terms of them.

We need a Fourier series representation of DT signals:

Just like in CT: how do we find the c_k ?

DT Fourier coefficients

Option 1: solve a linear system of equations.

This gives the system:

(This may look familiar!)

Option 2: obtain closed-form expressions like the CT case.

Leverage the following identity about complex numbers:

We will multiply on both sides, and sum.

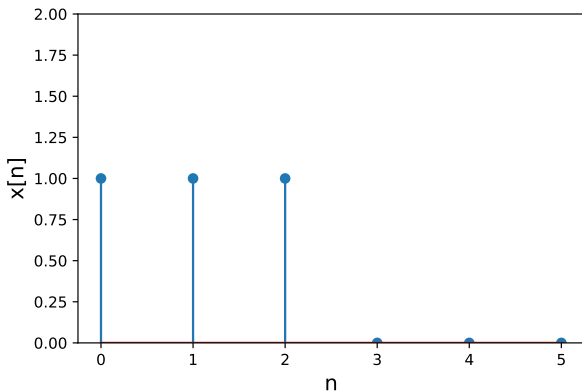
DT Fourier coefficients

So the DT synthesis and analysis equations are

and

Example: the DT square wave

Let's compute the Fourier coefficients of this signal:



Let's recap what we've done the previous three or so lectures.

Back in lectures 2 and 3, we talking about LTI systems and impulse response. We showed a couple of important things...

Any signal can be expressed as a combination of weighted, shifted impulses.

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

If we know what the system does to a unit impulse, i.e., the impulse response $h(t)$ or $h[n]$, we can learn what it does to any signal. This was the convolution integral and sum:

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

We now know many periodic signals, in both CT and DT, can be expressed as Fourier series (linear combinations of complex exponential signals):

System functions

And we know that these complex exponential signals are eigenfunctions of LTI systems:

In general, $H(s)$ is called the **system function**; when we are limiting to $H(j\omega)$ we term this the **frequency response**.

In other words, when we input a complex exponential signal at a particular frequency, how does the system respond?

Example: computing the frequency response

Consider an LTI system with impulse response:

What happens when we put in the following signal?

Example: computing the frequency response

Three steps:

1. Determine the Fourier coefficients of the signal
2. Determine the frequency response
3. Use 1 and 2 to obtain Fourier coefficients of the output signal

Example: computing the frequency response

Step 1: what are the Fourier coefficients of

Expand \sin and \cos as complex exponentials:

Then we see that:

Example: computing the frequency response

Step 2: what is the frequency response of the system with

Integrate!

Example: computing the frequency response

$$H(j\omega) = \int_{-\infty}^0 e^{(4-j\omega)\tau} d\tau + \int_0^{\infty} e^{-(4+j\omega)\tau} d\tau$$

Continue...

(What does this tell us about our system?)

Example: computing the frequency response

We have

Step 3: get the Fourier coefficients of the output signal.
In general, we know that

Example: computing the frequency response

Two sets of coefficients to deal with:

and

Example: computing the frequency response

So our original signal was

Our final signal is

Today's learning outcomes were:

- Compute the fundamental period and frequency of a DT signal
- Express DT periodic signals as Fourier series and compute the Fourier coefficients
- Compute the frequency response of a system, and determine how it responds to a complex exponential signal

What topics did you find unclear today?

For next time

We are going to learn how to *design* system functions to manipulate signals in certain ways:

- Frequency-shaping filters
- Frequency-selective filters

Action items:

1. Assignment 2 is due next Wednesday
2. Quiz 3 is on Tuesday

Recommended reading:

- From today's class: Oppenheim 1.3, 3.6-3.8
- For next class: Oppenheim 3.9