TECH 350: DSP Class VI: More Ways to Characterize LTI Systems





Tools for Describing Filters (So Far)

How can we describe filters?

- *First*, by their *frequency response*
 - (magnitude and phase response)
- Second, by how they are implemented
- (difference equations, underlying functions (Chebyshev, Butterworth, etc.), order)
 - *Third*, by contextualizing them vis-a-vis filter parameters/characteristics (cutoff frequency, resonance, etc.)

Example 1

System: y(n) = S(x(n)), where S: output = input * 0.5 + 0.5 **→** y(n) **x(n)** S

Input: 1 0.5 0 -0.5 -1



- where n = 0, 1, 2, 3, . . . are indices to input X and output Y



Output: 1 0.75 0.5 0.25 0





Input: 1 0 0 0 0 ...



The Impulse or Kronecker delta $\partial_{t,k}$, with k = 0 in this case

S → y(n)

Output: 1 0.5 0.25 0.125 0.0625...



The Impulse Response (IR) (AKA h(n)) of the system S...

but only for an impulse at t = 0.

What about at other times?



The Impulse Response

impulse is applied to the system input, often written as h[n]

What information does the IR give us?

What about for impulses at times other than k=0?

Answer: S is LTI, so we're all set! More specifically, we can define any input as time-shifted and scaled impulses, and because S is LTI, the output is equal to the sum of impulses, time-shifted and scales in an identical configuration (this is exceptionally important later).

Does the IR give us information about the full bandwidth of the digital system encapsulating S?

Answer: The impulse contains *all frequencies*, so the IR informs us how the system responds at all frequencies.

Impulse Response (IR): the output signal of a system S when an



Sound Typology (5)

(narrow in time, wide in frequency)



• The most complex sound is white noise (completely random)

plitude Amj

time

• Non-periodic sounds have no pitch and tend to have continuous spectra, e.g. a short pulse





Finite vs. Infinite Impulse Response

Finite Impulse Response (FIR)

A FIR filter has a fixed length IR (more explicitly, it becomes zero at some time t and stays that way!), A FIR filter of order N as an IR of length N+1 samples. FIR filters may have feedforward, but no

feedback.

Pros: Can have a flat (or variable) phase response, is more stable than IIR

Cons: is more computationally expensive

Example:

y(n) = 1/2(x(n) + x(n-1))

h(n) (as vector) = [0.5 0.5]



Finite vs. Infinite Impulse Response

Infinite Impulse Response (IIR)

An IIR filter has an IR that continues indefinitely. Many digital filters (any with feedback) are IIR. **Pros:** Less computationally expensive **Cons:** Less stability, nonlinear phase response

Example:

$$y(n) = 1/2(x(n) + y(n-1))$$
 h(n)

The biquadratic filter is also another IIR filter

(as vector) = [0.5 0.25 0.125 0.0625]

z-Transform

We define *Z* such that powers of Z correspond to the number of delays or advances in a difference equation.

One sample of delay = Z. Two samples of delay = Z^2 . One sample of advance = Z^{-1} . Two samples of advance = Z^{-2} .

Ex.:

If we're given a sequence $x(t) = [1 \ 2 \ 3 \ 4]$, its z-Transform, $X(Z) = 1 + 2Z + 3Z^2 + 4Z^3$

Ex.:

If we're given a sequence $x(t) = [1 \ 3 \ 0 \ -1 \ -5]$, its z-Transform, $X(Z) = 1 + 3Z - Z^3 - 5Z^4$

The Convolution Theorem

The convolution of f and g is written f*g.

An intuitive way to compute convolution by hand on 1-dimensional vectors is to think of it as a sliding window of multiplication and summation by a flipped 'kernel'. For example:

$$x_1 = (1, 5)$$

 $x_2 = (1, 2, 3)$

 $x_1 * x_2 = (1, 7, 13, 15)$

Convolution by hand (metaphor of sliding window):

The Convolution Theorem

$$x_1 = (1, 5)$$

 $x_2 = (1, 2, 3)$

z-Transforms of these: $x_1(Z) = 1 + 5Z$ $x_2(Z) = 1 + 2Z + 3Z^2$

Multiplying their z-Transforms together: $x_1(Z)x_2(Z) = 1 + 7Z + 13Z^2 + 15Z^3$

... which is the z-Transform of the convolution of x_1 and $X_2!!$

The Convolution Theorem

The Convolution Theorem: convolution in the time domain is equivalent to multiplication in the Z domain.

How does this help us?

Well, remember impulse responses of LTI systems?

A very important application of the convolution theorem: any signal x(n) that is input to an LTI system, the system's output y(x) is equal to the discrete convolution of the input signal x(n) and the system's impulse response (IR) h(n).

So, if $x_1 =$ input signal and $x_2 =$ a systems's impulse response, we can figure out its output!

For HW: Mini-Assignment II

z-Domain transforms

Practicing convolution of matrices

Some Octave commands + getting more knowledge of Complex Sinusoids

Calculating impulse responses

