

# ECE 4624: Meeting 7

## Time Domain Analysis of DT LTI Systems

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2025-09-18

Today we review how to perform time-domain analysis of DT LTI systems using convolution and review the concept of an Eigenfunction.

READING:

- ▶ PM 2.3 and 2.6
- ▶ ECE 2714 Supplementary Notes Chapters 9 and 13

Topics:

- ▶ Convolution
- ▶ DT time-domain analysis
- ▶ Correlation
- ▶ Eigen-Function for DT systems

## Review DT LTI systems and superposition property

If a DT system is LTI then the superposition property holds. Given a system where

$$x_i[n] \mapsto y_i[n] \quad \forall i$$

then

$$\sum_i a_i x_i[n] \mapsto \sum_i a_i y_i[n]$$

## Convolution Sum

Let  $h[n]$  be the DT impulse response. Then if the system is LTI the output for an input  $x[n]$  is given by

$$y[n] = \sum_{m=-\infty}^{\infty} x[m]h[n-m] = x[n] * h[n]$$

This is called the convolution sum.

## Graphical View of the Convolution Sum.

## Example

Find the convolution of two unit step functions.

## Example

Find the convolution if the signals

$$x_1[n] = \gamma^n u[n] \text{ for } \gamma \in \mathbb{R}$$

and

$$x_2[n] = u[n]$$

# Properties of DT Convolution

There are several useful properties of convolution. We do not prove these here, but it is not terribly difficult to do so. Given signals  $x_1[n]$ ,  $x_2[n]$ , and  $x_3[n]$ :

**Commutative Property** The ordering of the signals does not matter.

$$x_1[n] * x_2[n] = x_2[n] * x_1[n]$$

**Distributive Property** Convolution is distributed over addition.

$$x_1[n] * (x_2[n] + x_3[n]) = (x_1[n] * x_2[n]) + (x_1[n] * x_3[n])$$

**Associative Property** The order of convolution does not matter.

$$x_1[n] * (x_2[n] * x_3[n]) = (x_1[n] * x_2[n]) * x_3[n]$$

**Index Shift** Given  $x_3[n] = x_1[n] * x_2[n]$  then for index shifts  $m_1, m_2 \in \mathbb{R}$

$$x_1[n - m_1] * x_2[n - m_2] = x_3[n - m_1 - m_2]$$

**Multiplicative Scaling** Given  $x_3[n] = x_1[n] * x_2[n]$  then for constants  $a, b \in \mathbb{C}$

$$(a x_1[n]) * (b x_2[n]) = a b x_3[n]$$



## Example

Suppose an LTI system has an impulse response

$$h[n] = \delta[n + 1] - 2\delta[n] + \delta[n - 1]$$

What is the output if the input is

$$x[n] = \left(\frac{1}{2}\right)^{|n|}$$

## DT Convolution of Finite-Length Signals

- ▶ For finite-length signals, DT convolution gives us an algorithm to determine their convolution.
- ▶ Suppose the signal  $x_1$  is non-zero only over the interval  $[N_1, M_1]$ , and the signal  $x_2$  is non-zero only over the interval  $[N_2, M_2]$ .
- ▶ The *length* of the signals are  $L_1 = M_1 - N_1 + 1$  and  $L_2 = M_2 - N_2 + 1$  respectively.
- ▶ The non-zero terms of the convolution sum (when the signals overlap) is then the range  $[N_1 + N_2, M_1 + M_2]$  and the sum can be truncated as:

$$x_1[n] * x_2[n] = \sum_{m=N_1+N_2}^{M_1+M_2} x_1[m]x_2[n-m]$$

- ▶ It is common to shift both signals so that they both start at index 0 , zero-padding them both to have length  $L = L_1 + L_2 - 1$ .
- ▶ Then the convolution becomes

$$y = x_1 * x_2 = \sum_{m=0}^{L-1} x_1[m]x_2[L+n-m]$$

where the indexing of  $x_2$  is modulo the signal length.

- ▶ The resulting signal after convolution,  $y$ , is also of length  $L$ , and can then be shifted back to start at  $N_1 + N_2$ .

## Example

Compute the convolution of the DT signals  $\{1, -1, 1\}$  and  $\{1, 1, 1, 1\}$

## Code for Convolution of two finite length signals

For the example above:

```
const unsigned int L = 6; //L1=3, L2=4, so L=6
double x1[L] = {1., -1., 1., 0, 0, 0};
double x2[L] = {1., 1., 1., 1., 0, 0};
double y[L];

for(int n = 0; n < L; n++){
    double sum = 0.;
    for(int m = 0; m < L; m++){
        int idx = (L+n-m) % L;
        sum += x1[m]*x2[idx];
    }
    y[n] = sum;
}
```

## Correlation

An operation very similar to convolution is *correlation*. Given two signals  $x[n]$  and  $y[n]$

- ▶ their cross-correlation is

$$r_{xy}[l] = \sum_{n=-\infty}^{\infty} x[n]y[n-l]$$

where  $l$  is called the “lag”.

- ▶ note that

$$r_{yx}[l] = \sum_{n=-\infty}^{\infty} y[n]x[n-l] = r_{xy}[-l]$$

- ▶ if  $x = y$  we get the auto-correlation  $r_{xx}[l]$
- ▶ the normalized autocorrelation and cross-correlation are given by

$$\rho_{xx}[l] = \frac{r_{xx}[l]}{r_{xx}[0]}$$

$$\rho_{xy}[l] = \frac{r_{xy}[l]}{(r_{xx}[l] \cdot r_{yy}[l])^{\frac{1}{2}}}$$

## Correlation of input and output of LTI system

Given an LTI system with impulse response  $h[n]$  then

$$r_{yx}[l] = h[n] * r_{xx}[l]$$

## Applications of correlation

- ▶ “radar” equation
- ▶ The Fourier transform of the auto/cross correlation is the power spectral density. This forms the basis of working with random signals.

## Eigenfunction for DT-LTI Systems

Let  $x[n] = z^n$  for  $z \in \mathbb{C}$ , then  $y[n] = h[n] * x[n] = x[n] * h[n]$  and by the definition of DT convolution

$$\begin{aligned}y[n] &= \sum_{m=-\infty}^{\infty} h[m]x[n-m] \\&= \sum_{m=-\infty}^{\infty} h[m]z^{n-m} = \sum_{m=-\infty}^{\infty} h[m]z^n z^{-m} \\&= z^n \sum_{m=-\infty}^{\infty} h[m]z^{-m} \\&= z^n H(z)\end{aligned}$$

where  $H(z) = \sum_{m=-\infty}^{\infty} h[m]z^{-m}$  is the *Z Transform* of the impulse response,  $h[n]$ .

$H(z)$  is called the *transfer function* or *Eigenvalue* of the system and  $z^n$  is the *Eigenfunction* for DT LTI systems.