

EE 470 - Signals and Systems

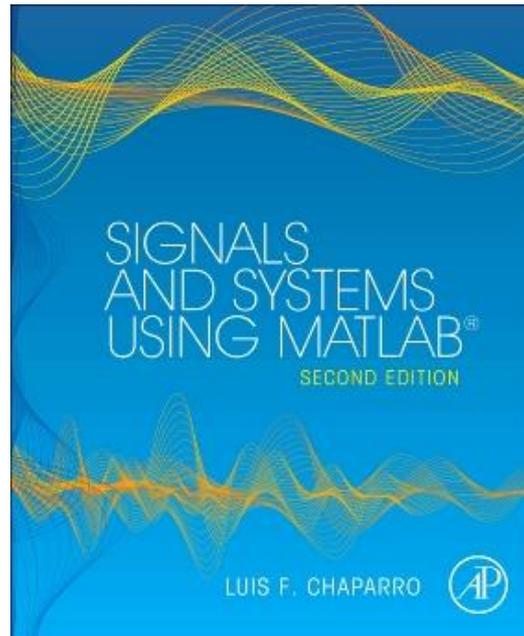
6. Discrete-Time Signals and Systems

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Textbook

Luis Chapparo, Signals and Systems Using Matlab, 2nd ed., Academic Press, 2015.



Discrete-Time Signals

- A discrete-time signal $x[n]$ can be thought of as a real- or complex-valued function of the integer sample index n :
 - For discrete-time signals the independent variable is an integer n , the sample index, and that the value of the signal at n , $x[n]$, is either real or complex
 - Signal is only defined at integer values n —no definition exists for values between the integers
- Example: sampled signal: $x(nT_s) = x(t)|_{t=nT_s}$

Discrete-Time Signals: Example

- Consider a sinusoidal signal:

$$x(t) = 3 \cos(2\pi t + \pi/4) \quad -\infty < t < \infty$$

Determine an appropriate sampling period T_s and obtain the discrete-time signal $x[n]$ corresponding to the largest allowed sampling period.

- *Solution:*

To sample $x(t)$ so that no information is lost, the Nyquist sampling rate condition indicates that the sampling

period should be: $T_s \leq \frac{\pi}{\Omega_{\max}} = \frac{\pi}{2\pi} = 0.5 \quad \Rightarrow \quad T_s^{\max} = 0.5$


$$x[n] = 3 \cos(2\pi t + \pi/4)|_{t=0.5n} = 3 \cos(\pi n + \pi/4) \quad -\infty < n < \infty$$

Periodic and Aperiodic Signals

A discrete-time signal $x[n]$ is *periodic* if

- It is defined for all possible values of n , $-\infty < n < \infty$.
- There is a positive integer N , the period of $x[n]$, such that

$$x[n + kN] = x[n]$$

for any integer k .

Periodic discrete-time sinusoids, of period N , are of the form

$$x[n] = A \cos\left(\frac{2\pi m}{N}n + \theta\right) \quad -\infty < n < \infty$$

where the discrete frequency is $\omega_0 = 2\pi m/N$ rad, for positive integers m and N , which are not divisible by each other, and θ is the phase angle.

Periodic and Aperiodic Signals: Example 1

- Consider the discrete sinusoids:

$$x_1[n] = 2 \cos(\pi n - \pi/3)$$

$$x_2[n] = 3 \sin(3\pi n + \pi/2) \quad -\infty < n < \infty$$

→ $\omega_1 = \pi = \frac{2\pi}{2}$ → $m = 1$ and $N = 2$, → periodic of period $N_1 = 2$

→ $\omega_2 = 3\pi = \frac{2\pi}{2} \cdot 3$ → $m = 3$ and $N = 2$ → periodic of period $N_2 = 2$

Periodic and Aperiodic Signals: Example 2

- Continuous-time sinusoids are always periodic but this is not true for discrete-time sinusoids
- Consider: $x[n] = \cos(n + \pi/4)$.

The sampled signal $x[n] = x(t)|_{t=nT_s} = \cos(n + \pi/4)$ has a discrete frequency $\omega = 1$ rad that cannot be expressed as $2\pi m/N$ for any integers m and N because π is an irrational number. So $x[n]$ is not periodic.

Since the frequency of the continuous-time signal $x(t)$ is $\Omega = 1$ (rad/sec), then the sampling period, according to the Nyquist sampling rate condition, should be

$$T_s \leq \frac{\pi}{\Omega} = \pi$$

and for the sampled signal $x(t)|_{t=nT_s} = \cos(nT_s + \pi/4)$ to be periodic of period N or

$$\cos((n + N)T_s + \pi/4) = \cos(nT_s + \pi/4) \quad \text{is necessary that} \quad NT_s = 2k\pi$$



$$T_s = 2k\pi/N \leq \pi$$

Sampling Analog Periodic Signal

When sampling an analog sinusoid

$$x(t) = A \cos(\Omega_0 t + \theta) \quad -\infty < t < \infty$$

of period $T_0 = 2\pi/\Omega_0$, $\Omega_0 > 0$, we obtain a *periodic discrete sinusoid*,

$$x[n] = A \cos(\Omega_0 T_s n + \theta) = A \cos\left(\frac{2\pi T_s}{T_0} n + \theta\right)$$

provided that

$$T_s \leq \frac{\pi}{\Omega_0} = \frac{T_0}{2}$$

$$\frac{T_s}{T_0} = \frac{m}{N}$$

Sum of Discrete-Time Period Signals

- The sum $z[n] = x[n] + y[n]$ of periodic signals $x[n]$ with period N_1 , and $y[n]$ with period N_2 is periodic if the ratio of periods of the summands is rational—that is,

$$\frac{N_2}{N_1} = \frac{p}{q}$$

- Here p and q are integers not divisible by each other
- If so, the period of $z[n]$ is $qN_2 = pN_1$
- Example: $z[n] = \sin(\pi n + 2) + \cos(2\pi n/3 + 1)$
 - $N_1 = 2$, $N_2 = 3$ and hence, sum is periodic with period 6
- Example: $z[n] = \sin(\pi n + 2) + \cos(2n/3 + 1)$
 - $N_1 = 1$, signal 2 is not periodic: sum is not periodic

Finite Energy and Finite Power Discrete-Time Signals

For a discrete-time signal $x[n]$, we have the following definitions:

$$\text{Energy: } \varepsilon_x = \sum_{n=-\infty}^{\infty} |x[n]|^2$$

$$\text{Power: } P_x = \lim_{N \rightarrow \infty} \frac{1}{2N+1} \sum_{n=-N}^N |x[n]|^2$$

- $x[n]$ is said to have *finite energy* or to be *square summable* if $\varepsilon_x < \infty$.
- $x[n]$ is called *absolutely summable* if

$$\sum_{n=-\infty}^{\infty} |x[n]| < \infty$$

- $x[n]$ is said to have *finite power* if $P_x < \infty$.

Time Shifting, Scaling, and Even/Odd Discrete-Time Signals

A discrete-time signal $x[n]$ is said to be

- *Delayed* by N (an integer) samples if $x[n - N]$ is $x[n]$ shifted to the right N samples.
- *Advanced* by M (an integer) samples if $x[n + M]$ is $x[n]$ shifted to the left M samples.
- *Reflected* if the variable n in $x[n]$ is negated (i.e., $x[-n]$).

Even and odd discrete-time signals are defined as

$$x[n] \text{ is even: } \Leftrightarrow x[n] = x[-n]$$

$$x[n] \text{ is odd: } \Leftrightarrow x[n] = -x[-n]$$

Any discrete-time signal $x[n]$ can be represented as the sum of an even and an odd component,

$$\begin{aligned} x[n] &= \frac{1}{2} \underbrace{(x[n] + x[-n])}_{x_e[n]} + \frac{1}{2} \underbrace{(x[n] - x[-n])}_{x_o[n]} \\ &= x_e[n] + x_o[n] \end{aligned}$$

Even/Odd: Example

Find the even and the odd components of the discrete-time signal

$$x[n] = \begin{cases} 4 - n & 0 \leq n \leq 4 \\ 0 & \text{otherwise} \end{cases}$$

$$x_e[n] = 0.5(x[n] + x[-n]) \Rightarrow x_e[n] = \begin{cases} 2 + 0.5n & -4 \leq n \leq -1 \\ 4 & n = 0 \\ 2 - 0.5n & 1 \leq n \leq 4 \\ 0 & \text{otherwise} \end{cases}$$

$$x_o[n] = 0.5(x[n] - x[-n]) \Rightarrow x_o[n] = \begin{cases} -2 - 0.5n & -4 \leq n \leq -1 \\ 0 & n = 0 \\ 2 - 0.5n & 1 \leq n \leq 4 \\ 0 & \text{otherwise} \end{cases}$$

Discrete-Time Unit-Step and Unit-Sample Signals

The unit-step $u[n]$ and the unit-sample $\delta[n]$ discrete-time signals are defined as

$$u[n] = \begin{cases} 1 & n \geq 0 \\ 0 & n < 0 \end{cases}$$

$$\delta[n] = \begin{cases} 1 & n = 0 \\ 0 & \text{otherwise} \end{cases}$$

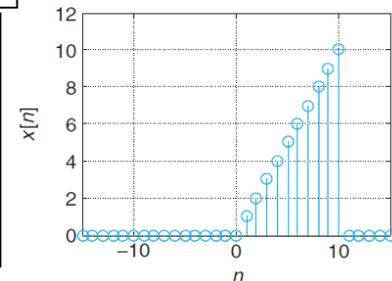
These two signals are related as follows:

$$\delta[n] = u[n] - u[n - 1]$$

$$u[n] = \sum_{k=0}^{\infty} \delta[n - k] = \sum_{m=-\infty}^n \delta[m]$$

Any discrete-time signal $x[n]$ is represented using unit-sample signals as

$$x[n] = \sum_{k=-\infty}^{\infty} x[k] \delta[n - k]$$



Discrete-Time Systems

- Just as with continuous-time systems, a discrete-time system is a transformation of a discrete-time input signal $x[n]$ into a discrete-time output signal $y[n]$:

$$y[n] = \mathcal{S}\{x[n]\}$$

A discrete-time system \mathcal{S} is said to be

- *Linear*: If for inputs $x[n]$ and $v[n]$ and constants a and b , it satisfies the following

- *Scaling*: $\mathcal{S}\{ax[n]\} = a\mathcal{S}\{x[n]\}$

- *Additivity*: $\mathcal{S}\{x[n] + v[n]\} = \mathcal{S}\{x[n]\} + \mathcal{S}\{v[n]\}$

or equivalently if *superposition* applies—that is,

$$\mathcal{S}\{ax[n] + bv[n]\} = a\mathcal{S}\{x[n]\} + b\mathcal{S}\{v[n]\}$$

- *Time-invariant*: If for an input $x[n]$ with a corresponding output $y[n] = \mathcal{S}\{x[n]\}$, the output corresponding to a delayed or advanced version of $x[n]$, $x[n \pm M]$, is $y[n \pm M] = \mathcal{S}\{x[n \pm M]\}$ for an integer M .

Recursive and Nonrecursive Discrete-Time Systems

- *Recursive system:*

$$y[n] = - \sum_{k=1}^{N-1} a_k y[n-k] + \sum_{m=0}^{M-1} b_m x[n-m] \quad n \geq 0$$

initial conditions $y[-k], k = 1, \dots, N-1$

This system is also called *infinite-impulse response* (IIR).

- *Nonrecursive system:*

$$y[n] = \sum_{m=0}^{M-1} b_m x[n-m]$$

This system is also called *finite-impulse response* (FIR).

Discrete-Time Systems: Example 1

- **Moving-average discrete filter:** 3rd-order moving-average filter (also called a smoother since it smoothes out the input signal) is an FIR filter for which the input $x[n]$ and the output $y[n]$ are related by:

$$y[n] = \frac{1}{3}(x[n] + x[n - 1] + x[n - 2])$$

- **Linearity: Yes**

$$\frac{1}{3}[(ax_1[n] + bx_2[n]) + (ax_1[n - 1] + bx_2[n - 1]) + (ax_1[n - 2] + bx_2[n - 2])] = ay_1[n] + by_2[n]$$

- **Time Invariance: Yes**

$$\begin{aligned} \frac{1}{3}(x_1[n] + x_1[n - 1] + x_1[n - 2]) &= \frac{1}{3}(x_1[n - N] + x_1[n - N - 1] + x_1[n - N - 2]) \\ &= y_1[n - N] \end{aligned}$$

Discrete-Time Systems: Example 2

- **Autoregressive discrete filter:** The recursive discrete-time system represented by the first-order difference equation (with initial condition $y[-1]$):

$$y[n] = ay[n - 1] + bx[n] \quad n \geq 0, \quad y[-1]$$

- **Autoregressive moving average filter:**

$$y[n] = 0.5y[n - 1] + x[n] + x[n - 1]$$

- Called the autoregressive moving average given that it is the combination of the two systems

Discrete-Time Systems Represented by Difference Equations

- General form:

$$y[n] = - \sum_{k=1}^{N-1} a_k y[n-k] + \sum_{m=0}^{M-1} b_m x[n-m] \quad n \geq 0$$

initial conditions $y[-k], k = 1, \dots, N-1$

- Just as in the continuous-time case, the system being represented by the difference equation is not LTI unless the initial conditions are zero and the input is causal
- Complete response of a system represented by the difference equation can be shown to be composed of a zero-input and a zero-state responses

$$y[n] = y_{zi}[n] + y_{zs}[n]$$

Discrete Convolution

- For LTI system with impulse response $h[n]$, starting from the generic representation of $x[n]$,

$$x[n] = \sum_{k=-\infty}^{\infty} x[k]\delta[n-k]$$

We can show that the output can be computed as:

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

$$\begin{aligned} [h * x][n] &= \sum_k x[k]h[n-k] = \sum_k x[n-k]h[k] \\ &= [x * h][n] \end{aligned}$$

Note: Convolution is a *linear operator*

Discrete Convolution: Example

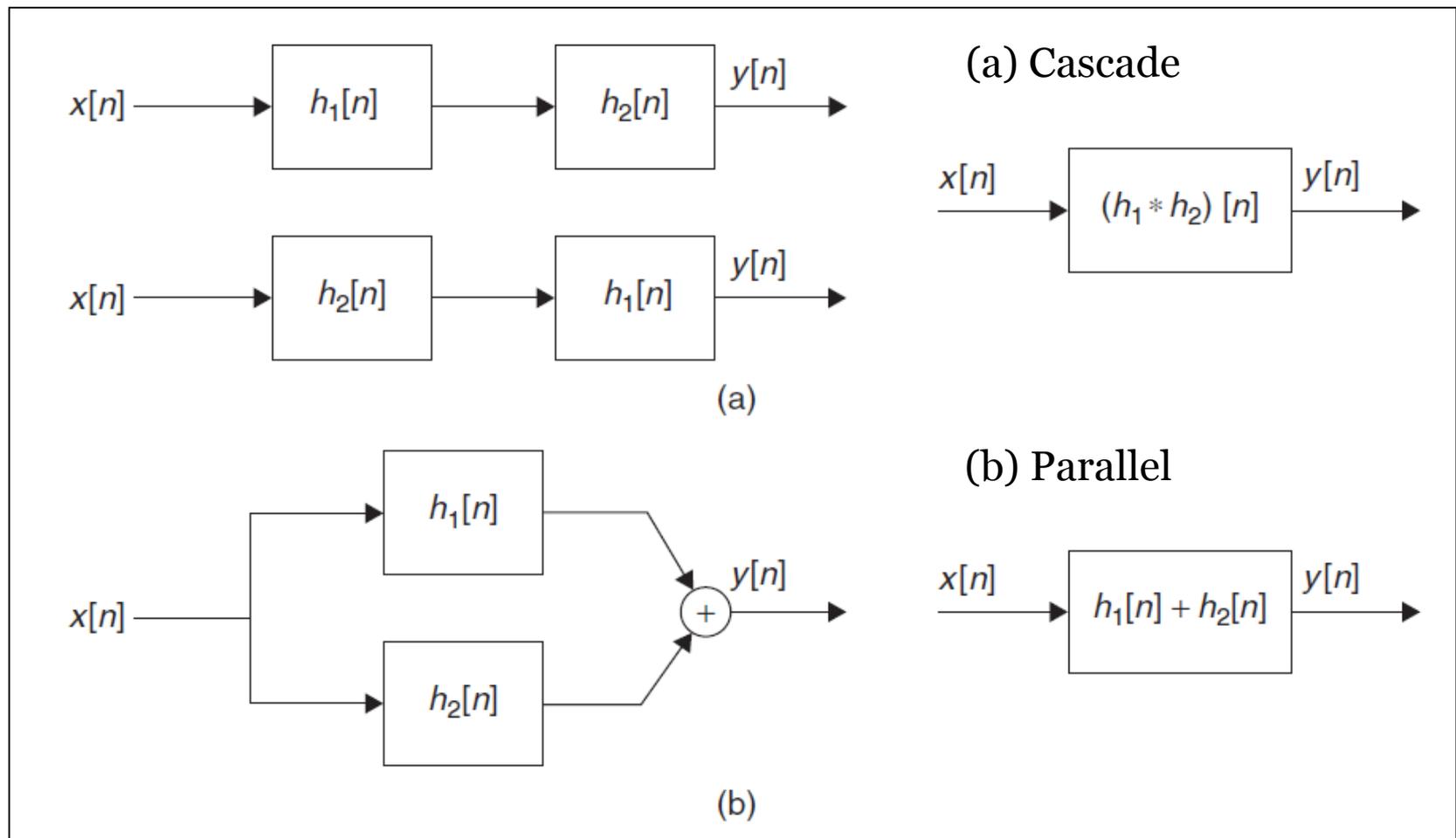
- The output of nonrecursive or FIR systems is the convolution sum of the input and the impulse response of the system:

$$y[n] = \sum_{k=0}^{N-1} b_k x[n-k]$$

- Impulse response is found when $x[n] = \delta[n]$.

$$h[n] = \sum_{k=0}^{N-1} b_k \delta[n-k] = b_0 \delta[n] + b_1 \delta[n-1] + \cdots + b_{N-1} \delta[n-(N-1)]$$

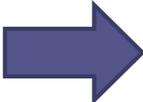
Cascade and Parallel Connections



Discrete-Time Systems: Example

- Find the impulse response and output for $x[n]=u[n]$ of a moving-averaging filter where the input is $x[n]$ and the output is $y[n]$:

$$y[n] = \frac{1}{3}(x[n] + x[n - 1] + x[n - 2])$$


$$h[n] = \frac{1}{3}(\delta[n] + \delta[n - 1] + \delta[n - 2])$$

$$y[0] = \frac{1}{3}(x[0] + x[-1] + x[-2]) = \frac{1}{3}x[0]$$

$$y[1] = \frac{1}{3}(x[1] + x[0] + x[-1]) = \frac{1}{3}(x[0] + x[1])$$

$$y[2] = \frac{1}{3}(x[2] + x[1] + x[0]) = \frac{1}{3}(x[0] + x[1] + x[2])$$

$$y[3] = \frac{1}{3}(x[3] + x[2] + x[1]) = \frac{1}{3}(x[1] + x[2] + x[3])$$

...

Thus, if $x[n]=u[n]$, then:

$$y[0] = 1/3$$

$$y[1] = 2/3$$

$$y[n] = 1 \quad \text{for } n \geq 2$$

Causality of Discrete-Time Systems

- A discrete-time system S is **causal** if:
 - Whenever the input $x[n]=0$, and there are no initial conditions, the output is $y[n]=0$.
 - The output $y[n]$ does not depend on future inputs.

- An LTI discrete-time system is *causal* if the impulse response of the system is such that

$$h[n] = 0 \quad n < 0$$

- A signal $x[n]$ is said to be *causal* if

$$x[n] = 0 \quad n < 0$$

- For a causal LTI discrete-time system with a causal input $x[n]$ its output $y[n]$ is given by

$$y[n] = \sum_{k=0}^n x[k]h[n-k] \quad n \geq 0$$

Causality: Examples

- Consider the system defined by,

$$y[n] = x^2[n]$$

- Nonlinear, time invariant and Causal
- Consider the moving average system defined by,

$$y[n] = \frac{1}{3}(x[n+1] + x[n] + x[n-1]).$$

- LTI and Non-Causal

Stability of Discrete-Time Systems

- Bounded-Input Bounded-Output (BIBO) Stability
- An LTI discrete-time system is said to be BIBO stable if its impulse response $h[n]$ is absolutely summable:

$$\sum_k |h[k]| < \infty$$

- Notes:
 - Nonrecursive or FIR systems are BIBO stable. Indeed, the impulse response of such a system is of finite length and thus absolutely summable.
 - For a recursive or IIR system represented by a difference equation, to establish stability we need to find the system impulse response $h[n]$ and determine whether it is absolutely summable or not.

Stability: Example

- Consider an autoregressive system

$$y[n] = 0.5y[n-1] + x[n]$$

Determine if the system is BIBO stable.

$$h[n] = 0.5^n u[n] \quad \Rightarrow \quad \sum_{n=-\infty}^{\infty} |h[n]| = \sum_{n=0}^{\infty} 0.5^n = \frac{1}{1-0.5} = 2$$

 System is BIBO stable

Problem Assignments

- Chapter 9 of Chaparro's textbook
- Assigned Problem Set #6