# Signals and Systems Using MATLAB

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Chapter 9 - Discrete-time Signals and Systems

# What is in this chapter?

- . Discrete-time signals

  - Basic discrete-time signals
- Discrete-time LTI systems
  - Recursive and non-recursive systems
  - Difference representation of systems
- Convolution sum
  - Causality and stability

#### 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:

$$x[.]: \mathcal{I} \to \mathcal{R} \quad (\mathcal{C})$$
 $n \quad x[n]$ 

- $x(nT_s)$  depends on n once  $T_s$  is known and is only defined at  $nT_s$
- Most discrete-time signals are obtained by sampling continuous-time signals, but there are inherently discrete signals

# Examples

(a) Sampling  $x(t) = 3\cos(2\pi t + \pi/4)$   $-\infty < t < \infty$  using  $T_s$  satisfying Nyquist sampling rate condition

$$T_s \le \frac{\pi}{\Omega_{max}} = \frac{\pi}{2\pi} = 0.5$$

for largest allowed sampling period  $T_s = 0.5$ 

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

(b) Fibonacci sequence  $\{x[n]\}$ 

$$x[n] = x[n-1] + x[n-2]$$
  $n \ge 2$   
 $x[0] = 0$   
 $x[1] = 1$ 

it has been used to model different biological systems

$$x[2] = 1 + 0 = 1$$
  
 $x[3] = 1 + 1 = 2$   
 $x[4] = 2 + 1 = 3$   
 $x[5] = 3 + 2 = 5$   
 $\vdots$   
 $x[n] = x[n-1] + x[n-2]$ 

Sequence is purely discrete as it is not related to a continuous—time signal Periodic and Aperiodic Signals

x[n] is periodic if

- ullet it is defined for all possible values of  $n, -\infty < n < \infty,$  and
- ullet there is a positive integer N, the period of x[n], such that

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

for any integer k.

Aperiodic signals are non-periodic

Sum z[n] = x[n] + y[n] of periodic signals x[n] of period  $N_1$ , and y[n] of period  $N_2$  is periodic if

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

i.e., p and q are integers, and not divisible by each other. The period of z[n] is  $qN_2 = pN_1$ 

Example z[n] = v[n] + w[n] + y[n], v[n], w[n] and y[n] periodic of periods  $N_1 = 2$ ,  $N_2 = 3$  and  $N_3 = 4$ , respectively. Determine if z[n] is periodic, and if so its period

$$x[n] = v[n] + w[n]$$
, so that  $z[n] = x[n] + y[n]$ 

x[n] is periodic since  $N_2/N_1 = 3/2$  is a rational number, and its period is  $N_4 = 3N_1 = 2N_2 = 6$ 

z[n] is also periodic since

$$\frac{N_4}{N_3} = \frac{6}{4} = \frac{3}{2}$$

is rational. Its period is  $N = 2N_4 = 3N_3 = 12$ , i.e.,

$$z[n+12] = v[n+6N_1] + w[n+4N_2] + y[n+3N_3] = v[n] + w[n] + y[n] = z[n]$$

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

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

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  $\theta$  is the phase angle.

$$x[n+kN] = A\cos\left(\frac{2\pi m}{N}(n+kN) + \theta\right) = A\cos\left(\frac{2\pi m}{N}n + 2\pi mk + \theta\right) = x[n]$$

When sampling

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

$$\frac{T_s}{T_0} = \frac{m}{N}$$
,  $N, m > 0$  and not divisible by each other

To avoid frequency aliasing also

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

**Example** Is  $x[n] = \cos(n + \pi/4)$  obtained by sampling  $x(t) = \cos(t + \pi/4)$  with  $T_s = 1$  periodic?

if so, indicate its period. Otherwise, find sampling period satisfying Nyquist and when used in sampling x(t) results in periodic signal

 $x[n] = x(t)|_{t=nT_s} = \cos(n + \pi/4)$ , is not periodic,  $\omega = 1$  (rad) cannot be expressed as  $2\pi m/N$  for integers m and N ( $\pi$  is irrational)

x(t) has frequency  $\Omega = 1$  then Nyquist requires  $T_s \leq \frac{\pi}{\Omega} = \pi$  for  $x(t)|_{t=nT_s} = \cos(nT_s + \pi/4)$  periodic of period N then

 $\cos((n+N)T_s+\pi/4) = \cos(nT_s+\pi/4)$  is necessary that  $NT_s = 2k\pi$ , k integer

 $T_s = 2k\pi/N \le \pi$  satisfies Nyquist sampling condition and insures the periodicity of  $x[nT_s)$ 

If period N = 10, then  $T_s = 0.2k\pi$ , for k chosen so Nyquist condition is satisfied, i.e.,

$$0 < T_s = k\pi/5 \le \pi$$
 so that  $0 < k \le 5$ 

Choose k=1 and 3 so that N and k are not divisible by each other If k=2, and 4 would give 5 as the period, and k=5 would give a period of 2 instead of 10

If k=1 then  $T_s=0.2\pi$  satisfies Nyquist sampling rate condition, and

$$x[n] = \cos(0.2n\pi + \pi/4) = \cos\left(\frac{2\pi}{10}n + \frac{\pi}{4}\right)$$

is periodic of period 10. The same for k=3

# Finite Energy and Finite Power Discrete-time Signals

For discrete-time signal x[n]:

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

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

- x[n] is finite energy or square summable if  $\varepsilon_x < \infty$ .
- x[n] is absolutely summable if

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

• x[n] is finite power if  $P_x < \infty$ .

Example "Causal" sinusoid,

$$x[n] = 2\cos(\Omega_0 t - \pi/4)u(t)|_{t=0.1n} = \begin{cases} 2\cos(0.1\Omega_0 n - \pi/4) & n \ge 0\\ 0 & \text{otherwise} \end{cases}$$

Does x[n] have finite energy, finite-power as compared with x(t) when  $\Omega_0 = \pi$ and when  $\Omega_0 = 3.2 \text{ rad/sec}$  (an upper approximation of  $\pi$ )?

x(t) has infinite energy, and so does x[n], for all  $\Omega_0$ 

$$\varepsilon_x = \sum_{n=-\infty}^{\infty} x[n]^2 = \sum_{n=0}^{\infty} 4\cos^2(0.1\Omega_0 n - \pi/4) \to \infty$$

x(t) and x[n] have finite power

(i)  $\Omega_0 = \pi$ ,  $x[n] = 2\cos(\pi n/10 - \pi/4) = 2\cos(2\pi n/20 - \pi/4)$  for  $n \ge 0$  and zero otherwise

x[n] repeats every  $N_0 = 20$  samples for  $n \geq 0$ , so

$$P_{x} = \lim_{N \to \infty} \frac{1}{2N+1} \sum_{n=-N}^{N} |x[n]|^{2} = \lim_{N \to \infty} \frac{1}{2N+1} \sum_{n=0}^{N} |x[n]|^{2}$$

$$= \lim_{N \to \infty} \frac{1}{2N+1} N \underbrace{\left[\frac{1}{N_{0}} \sum_{n=0}^{N_{0}-1} |x[n]|^{2}\right]}_{\text{power of period, } n \ge 0} = \frac{1}{2N_{0}} \sum_{n=0}^{N_{0}-1} |x[n]|^{2} < \infty$$

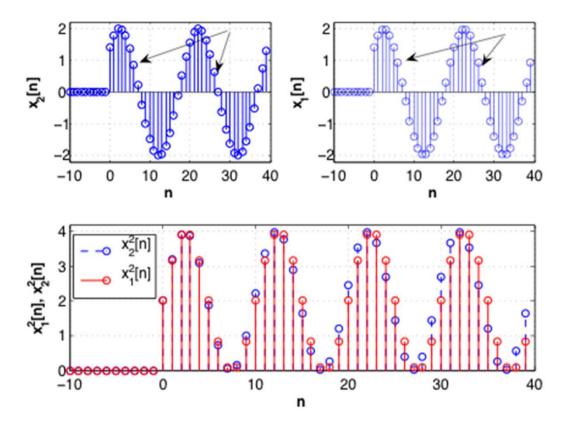
$$= \frac{4}{40} 0.5 \left[\sum_{n=0}^{10} 1 + \sum_{n=0}^{19} \cos(0.2\pi n - \pi/2)\right] = \frac{2}{40} [20 + 0] = 1$$

(ii)  $\Omega_0 = 3.2$ ,  $x[n] = 2\cos(3.2n/10 - \pi/4)$  for  $n \ge 0$  and zero otherwise; it does not repeat periodically for  $n \ge 0$ 

3.2/10 cannot be expressed as  $2\pi m/N$ , no close form for the power

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

and conjecture that because x(t) has finite power, so would x[n]



Even and 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].

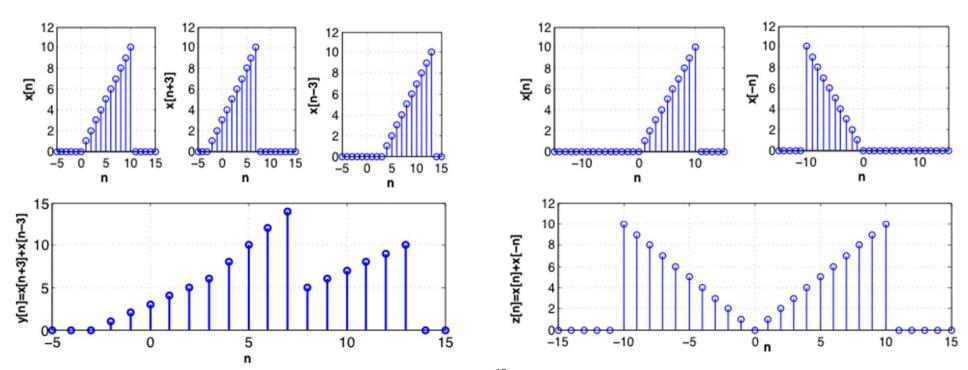
Even and odd discrete-time signals are defined as

$$x[n]$$
 is **even**:  $\Leftrightarrow$   $x[n] = x[-n]$   
 $x[n]$  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 components

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

$$= x_e[n] + x_o[n]$$



# Example Find even and odd components of

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

$$x_{e}[n] = 0.5(x[n] + x[-n])$$

$$= \begin{cases} 2 + 0.5n & -4 \le n \le -1 \\ 4 & n = 0 \\ 2 - 0.5n & 1 \le n \le 4 \\ 0 & \text{otherwise} \end{cases}$$

$$x_o[n] = 0.5(x[n] - x[-n])$$

$$= \begin{cases} -2 - 0.5n & -4 \le n \le -1 \\ 0 & n = 0 \\ 2 - 0.5n & 1 \le n \le 4 \\ 0 & \text{otherwise} \end{cases}$$

# Basic Discrete-time Signals

# Discrete-time Complex Exponential

Given complex numbers  $A = |A|e^{j\theta}$  and  $\alpha = |\alpha|e^{j\omega_0}$ , a discrete-time complex exponential is a signal of the form

$$x[n] = A\alpha^{n}$$

$$= |A||\alpha|^{n}e^{j(\omega_{0}n+\theta)}$$

$$= |A||\alpha|^{n} \left[\cos(\omega_{0}n+\theta) + j\sin(\omega_{0}n+\theta)\right]$$

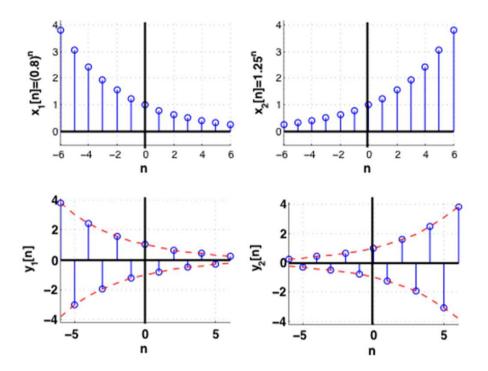
where  $\omega_0$  is a discrete frequency in radians.

1. Discrete-time vs continuous-time complex exponentials (for simplicity we let A be real) using as sampling period  $T_s$ 

Sampling 
$$x(t) = Ae^{(-a+j\Omega_0)t}$$
,  $A$  real  $x[n] = x(nT_s) = Ae^{(-anT_s+j\Omega_0nT_s)} = A(e^{-aT_s})^n e^{j(\Omega_0T_s)n} = A\alpha^n e^{j\omega_0n}$   $\alpha = e^{-aT_s}$ ,  $\omega_0 = \Omega_0T_s$ 

2. For  $\alpha > 0$  the real exponential

$$x[n] = (-\alpha)^n = (-1)^n \alpha^n = \alpha^n \cos(\pi n)$$



Real exponential  $x_1[n]=0.8^n$ ,  $x_2[n]=1.25^n$  (top) and modulated  $y_1[n]=x_1[n]\cos(\pi n)$  and  $y_2[n]=x_2[n]\cos(\pi n)$ 

**Example** Determine a > 0,  $\Omega_0$  and  $T_s$  for

$$x(t) = e^{-at} \cos(\Omega_0 t) u(t)$$

that permit us to obtain a discrete-time signal

$$y[n] = \alpha^n \cos(\omega_0 n)$$
  $n \ge 0$ 

and zero otherwise by sampling it. If  $\alpha = 0.9$  and  $\omega_0 = \pi/2$ , find a,  $\Omega_0$  and  $T_s$  that will permit us to obtain y[n] from x(t) by sampling

Comparing 
$$x(nT_s)$$
 with  $y[n]$   
 $\alpha = e^{-aT_s}$   
 $\omega_0 = \Omega_0 T_s$ 

No unique solution. According to Nyquist condition

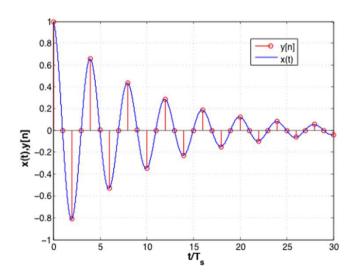
$$T_s \le \frac{\pi}{\Omega_{max}}$$

If  $\Omega_{max} = N\Omega_0$  for  $N \geq 2$  (signal is not band-limited so maximum frequency is not known, we are assuming it is a multiple of  $\Omega_0$ )

If 
$$T_s = \pi/N\Omega_0$$
  
 $\alpha = e^{-a\pi/N\Omega_0}$   
 $\omega_0 = \Omega_0 \pi/N\Omega_0 = \pi/N$ 

For  $\alpha = 0.9$ ,  $\omega_0 = \pi/2$ , we have

$$\begin{split} N &= 2 \\ a &= -\frac{2\Omega_0}{\pi} \log 0.9 \\ \text{If } \Omega_0 &= 2\pi \quad \Rightarrow \quad a = -4 \log 0.9, \quad T_s = 0.25 \end{split}$$



#### Discrete-time Sinuoids

$$x[n] = A\cos(\omega_0 n + \theta) = A\sin(\omega_0 n + \theta + \pi/2)$$
  $-\infty < n < \infty$ 

If discrete frequency  $w_0 = 2\pi m/N$  (rad), for integers m and N > 0 which are not divisible x[n] is periodic, oherwise not

Discrete frequency  $\omega=\pi/2, 5\pi/2, 9\pi/2\cdots=\pi/2$   $\omega=\pi, 3\pi, 5\pi\cdots=\pi=-\pi$   $\omega=0, 2\pi, 4\pi, \cdots=0$   $\omega=3\pi/2, 7\pi/2, 11\pi/2, \cdots=-\pi/2$ 

$$x_{1}[n] = \sin(0.1\pi n) = \sin\left(\frac{2\pi}{20}n\right)$$

$$x_{2}[n] = \sin(0.2\pi n) = \sin\left(\frac{2\pi}{10}n\right)$$

$$x_{3}[n] = \sin(0.6\pi n) = \sin\left(\frac{2\pi}{10}3n\right)$$

$$x_{4}[n] = \sin(0.7\pi n) = \sin\left(\frac{2\pi}{20}7n\right)$$

# 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

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

These two signals are related as follows

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

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

Generic representation of discrete-time signals

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

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

**Example** A train of triangular, discrete-time pulses t[n] of period N = 11 has a period

$$\tau[n] = \begin{cases} n & 0 \le n \le 5 \\ -n+10 & 6 \le n \le 10 \\ 0 & \text{otherwise} \end{cases}.$$

Find then an expression for its finite difference d[n] = t[n] - t[n-1]

$$t[n] = \dots + \tau[n+11] + \tau[n] + \tau[n-11] + \dots = \sum_{k=-\infty}^{\infty} \tau[n-11k]$$

The finite difference d[n] is then

$$d[n] = t[n] - t[n-1]$$

$$= \sum_{k=-\infty}^{\infty} (\tau[n-11k] - \tau[n-1-11k])$$

The signal d[n] is also periodic of the same period N = 11 as t[n]. If we let

$$s[n] = \tau[n] - \tau[n-1] = \begin{cases} 1 & 0 \le n \le 5 \\ -1 & 6 \le n \le 10 \\ 0 & \text{otherwise} \end{cases}$$

then

$$d[n] = \sum_{k=-\infty}^{\infty} s[n-11k]$$

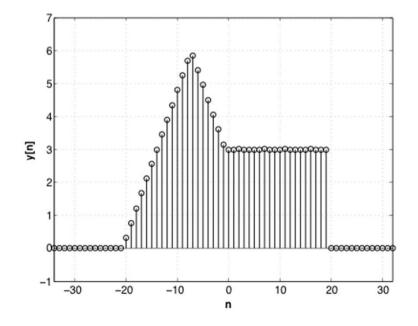
**Example** Express as function of n the sampled signal

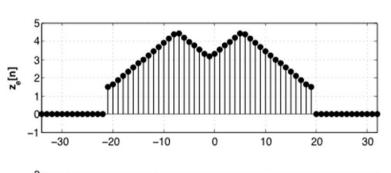
$$y[n] = 3r(t+3) - 6r(t+1) + 3r(t) - 3u(t-3)|_{t=0.15n}$$

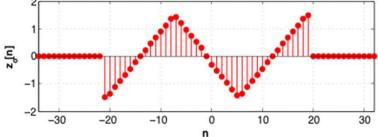
Use MATLAB to find and plot its even and odd components

$$y(t) = \begin{cases} 0 & t < -3 \\ 3r(t+3) = 3t+9 & -3 \le t < -1 \\ 3t+9-6r(t+1) = -3t+3 & -1 \le t < 0 \\ -3t+3+3r(t) = 3 & 0 \le t < 3 \\ 3-3=0 & t \ge 3 \end{cases}$$

$$y[n] = \begin{cases} 0 & n \le -21 \\ 0.45n+9 & -20 \le n \le -6 \\ -0.45n+3 & -7 \le n \le 0 \\ 3 & 1 \le n \le 19 \\ 0 & n \ge 20 \end{cases}$$







# Discrete-time Systems

 $Dynamic\ system$ 

$$y[n] = S\{x[n]\}$$
  $x[n]$  input,  $y[n]$  output

that is

- Linearity
- Time-invariance
- Stability
- Causality

A discrete-time system S is said to be

- Linear: if for inputs x[n] and v[n], and constants a and b, it satisfies the following
  - Scaling :  $S\{ax[n]\} = aS\{x[n]\}$
  - Additivity:  $S{x[n] + v[n]} = S{x[n]} + S{v[n]}$

or equivalently if superposition applies, i.e.,

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

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

## Example A Square-root Computation System

Difference equation, with some initial condition y[0], can be used to find the square root of  $\alpha$ :

$$y[n] = 0.5 \left[ y[n-1] + \frac{\alpha}{y[n-1]} \right] \qquad n > 0$$

Find recursively solution of this difference equation for  $\alpha=4$  and 2. Is system linear?

Difference equation is first-order, non-linear

Recursive solution

$$y[1] = 0.5 \left[ y[0] + \frac{\alpha}{y[0]} \right]$$

$$y[2] = 0.5 \left[ y[1] + \frac{\alpha}{y[1]} \right]$$

$$y[3] = 0.5 \left[ y[2] + \frac{\alpha}{y[2]} \right]$$
:

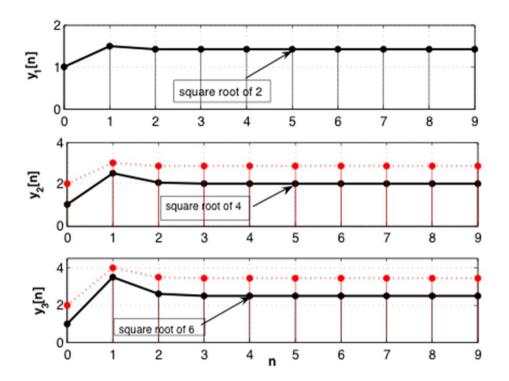
If y[0] = 1, and  $\alpha = 4$  (i.e., we wish to find the square root of 4),

$$y[0] = 1$$

$$y[1] = 0.5 \left[ 1 + \frac{4}{1} \right] = 2.5$$

$$y[2] = 0.5 \left[ 2.5 + \frac{4}{2.5} \right] = 2.05$$
:

converging to 2. When  $n \to \infty$  then y[n] = y[n-1] = Y, so Y = 0.5Y + 0.5(4/Y) or  $Y = \sqrt{4} = 2$ 



Non-linear system: square root of 2 (top), square root of 4 compared with twice the square root of 2, (bottom) sum of previous responses with response when computing square root of 2+4. Middle figure shows scaling does not hold and the bottom figure that additivity does not hold, either. System is non-linear.

# Recursive and Non-recursive Discrete-time Systems

Depending on the relation between the input x[n] and the output y[n] two types of discrete-time systems of interest are:

#### • Recursive system

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

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

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

# • Non-recursive system

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

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

Example Moving-average Discrete Filter

Show that the third-order moving-average FIR filter (also called a smoother )

$$y[n] = \frac{1}{3}(x[n] + x[n-1] + x[n-2])$$
 with input  $x[n]$  and output  $y[n]$  is LTI

<u>Linearity</u> — Let input be  $ax_1[n] + bx_2[n]$ , and  $\{y_i[n], i = 1, 2\}$  are the corresponding outputs to  $\{x_i[n], i = 1, 2\}$ , then

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

thus linear.

<u>Time invariance</u> — If input is  $x_1[n] = x[n-N]$  the corresponding output to it is

$$\frac{1}{3}(x_1[n] + x_1[n-1] + x_1[n-2]) = \frac{1}{3}(x[n-N] + x[n-N-1] + x[n-N-2])$$
$$= y[n-N]$$

i.e., the system is time-invariant.

Recursive systems are represented by difference equation

$$y[n] = -\sum_{k=1}^{N-1} a_k y[n-k] + \sum_{m=1}^{M-1} b_m x[n-m], \qquad n \ge 0, \text{ ICs } y[-k], \ k = 1, \dots, N-1$$

- Difference equation characterizes dynamics of discrete systems
- Difference equation could be approximation of differential equation representing continuous-time system
- Complete response

$$y[n] = y_{zi}[n] + y_{zs}[n]$$
,  $y_{zi}[n]$  zero-input response,  $y_{zs}[n]$  zero-state response  
=  $y_t[n] + y_{ss}[n]$ ,  $y_t[n]$  transient response,  $y_{ss}[n]$  steady-state response

#### Convolution Sum

Let h[n] be the impulse response of an LTI discrete-time system, or the output of the system corresponding to an impulse  $\delta[n]$  as input, and initial conditions (if needed) equal to zero.

Using the generic representation of the input x[n] of the LTI system

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

the output of the system is given by either of the following two forms of the convolution sum:

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

#### Remarks

 The output of an 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  $h[n] = \sum_{k=0}^{N-1} b_k \delta[n-k] = b_0 \delta[n] + b_1 \delta[n-1] + \dots + b_{N-1} \delta[n-(N-1)]$  
$$h[n] = b_n, \quad n = 0, \dots, N-1 \quad \Rightarrow \quad y[n] = \sum_{k=0}^{N-1} h[k] x[n-k]$$

Convolution sum as an operator is linear and commutative

$$[h*x][n] = \sum_{k} x[k]h[n-k] = \sum_{k} x[n-k]h[k]$$
$$= [x*h][n]$$

• Just as with analog systems, when conecting two LTI discrete-time systems (with impulse responses  $h_1[n]$  and  $h_2[n]$ ) in series or in parallel, their respective impulse responses are given by  $[h_1 * h_2][n]$  and  $h_1[n] + h_2[n]$ 

Example Moving-averaging filter

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

Find the impulse response h[n] of this filter. Then for

- 1. x[n] = u[n], find y[n] using the input-output relation and the convolution sum.
- 2.  $x[n] = A\cos(2\pi n/N)u[n]$ , determine the values of A, and N, so that the steady state response of the filter is zero
- (1)  $x[n] = \delta[n], y[n] = h[n],$  no initial conditions are needed

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

The convolution sum coincides with the input/output equation. This holds for any FIR filter.

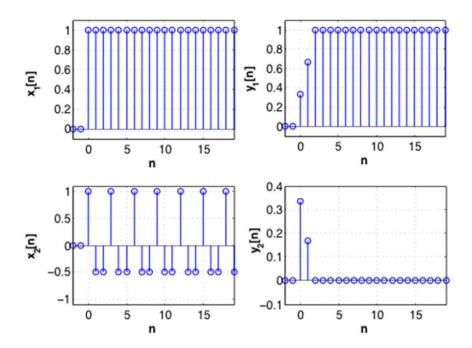
$$y[n] = 0, n < 0$$

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

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

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

If x[n] = u[n] then we have that y[0] = 1/3, y[1] = 2/3 and y[n] = 1 for  $n \ge 2$ . (2) For  $n \ge 2$ , y[n] is the average of the present and past two values of the input. If  $x[n] = A\cos(2\pi n/N)$ , if we let N = 3, and A be any real value the input repeats every 3 samples and the local average of 3 of its values is zero, giving y[n] = 0 for  $n \ge 2$ 

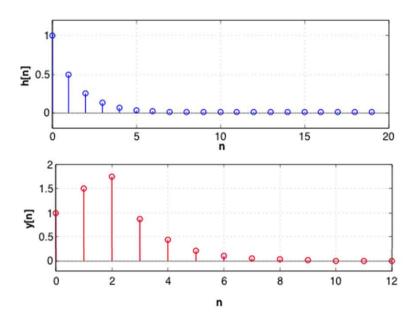


# Example. Autoregressive system

$$y[n] = 0.5y[n-1] + x[n] \qquad n \ge 0$$

Find the impulse response h[n] of the system and then compute the response of the system to x[n] = u[n] - u[n-3] using the convolution sum. Verify results with MATLAB.

```
a=[1 -0.5];b=1; % coefficients of the difference equation
d=[1 zeros(1,99)]; % approximate delta function
h=filter(b,a,d); % impulse response
x=[ones(1,3) zeros(1,10)]; % input
y=filter(b,a,x); % output from filter function
y1=conv(h,x); y1=y1(1:length(y)) % output from conv
```



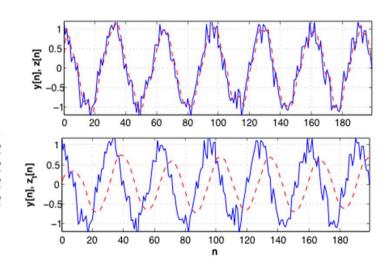
# Linear and Non-linear Filtering

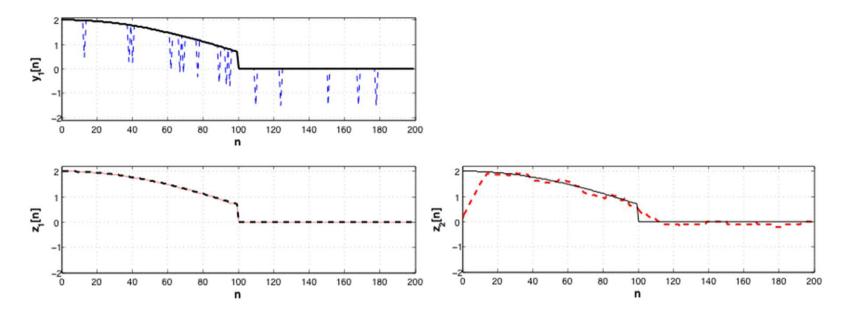
# Linear filtering

 $y[n] = x[n] + \eta[n]$ , where x[n] desired signal and  $\eta[n]$  noise

$$z[n] = \frac{1}{M} \sum_{k=0}^{M-1} y[n-k]$$
 averaging filter

Averaging filtering with filters of order M=3 (top figure), and of order M=15 result (bottom figure) used to get rid of Gaussian noise added to a sinusoid  $x[n]=\cos(\pi n/16)$ . Solid line corresponds to the noisy signal, while the dashed line is for the filtered signal. The filtered signal is very much like the noisy signal (see top figure) when M=3 is the order of the filter, while the filtered signal looks like the sinusoid, but shifted, when M=15.





Non-linear 5<sup>th</sup>-order median filtering (bottom left) versus linear 15<sup>th</sup>-order averager (bottom right) corresponding to the noisy signal (dashed line) and clean signal (solid line) on top. Clean signal (solid line) is superposed on de-noised signal (dashed line) in the bottom figures.

# Causality of Discrete-time Systems

- Real-time processing requires causality
- If data is stored no causality is needed

# 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

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

• A signal x[n] is said to be causal if

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

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

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

where the lower limit of the sum depends on the input causality, x[k] = 0 for k < 0, and the upper limit on the causality of the system, h[n-k] = 0 for n-k < 0 or k > n.

BIBO Stability of LTI Discrete-time Systems

LTI system represented by convolution sum

$$|y[n]| \leq \left|\sum_{k=-\infty}^{\infty} x[n-k]h[k]\right| \leq \sum_{k=-\infty}^{\infty} |x[n-k]||h[k]| \leq M \sum_{k=-\infty}^{\infty} |h[k]| \leq ML < \infty$$

An LTI discrete-time system is said to be BIBO stable if its impulse response h[n] is absolutely summable

$$\sum_{k} |h[k]| < \infty$$

Example Deconvolution: assume the input x[n] and the output y[n] of a causal LTI system are given, find equations to compute recursively the impulse response h[n] of the system. Consider finding the impulse response h[n] of a causal LTI system with input x[n] = u[n] and output  $y[n] = \delta[n]$ . Use the MATLAB deconv to find h[n].

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

$$h[n] = \frac{1}{x[0]} \left[ y[n] - \sum_{m=1}^{n} h[n-m]x[m] \right]$$

$$h[0] = \frac{1}{x[0]} y[0]$$

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

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

$$\vdots$$

When 
$$y[n] = \delta[n]$$
 and  $x[n] = u[n]$ 

$$h[0] = \frac{1}{x[0]}y[0] = 1$$

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

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

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

$$\vdots$$

and in general  $h[n] = \delta[n] - \delta[n-1]$ .

When using deconv make sure that the length of y[n] is always larger than that of x[n]

```
% Deconvolution
clear all
x=ones(1,100);
y=[1 zeros(1,100)]; % (a) correct h
% y=[1 zeros(1,99)]; % (b) incorrect h
[h,r]=deconv(y,x)
```

#### Remarks

- Non-recursive or FIR systems are BIBO stable, as their impulse responses are of finite length and absolutely summable.
- 2. For a recursive or IIR system represented by a difference equation, to established stability we need to find the system impulse response h[n] and determine whether it is absolutely summable or not.
- 3. A much simpler way to test the stability of an IIR system will be based on the location of the poles of the Z-transform of h[n], as we will see in the next chapter.

Example Consider an autoregressive system

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

Determine if the system is BIBO stable.

The impulse response of the system is  $h[n] = 0.5^n u[n]$ , checking the BIBO stability condition

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

thus the system is BIBO stable.

# What have we accomplished?

- Obtained similar theory for discrete- and continuous-time signals and systems with distinct differences
  - . Sampling period determines radian frequency of discrete signals
  - Discrete frequency is finite but circular
  - Discrete sinusoids not necessarily periodic

# Where do we go from here?

- . Z-transform analysis
  - . Z-transform relation to Laplace transform
    - Discrete-time Fourier transform and relation with Z-transform