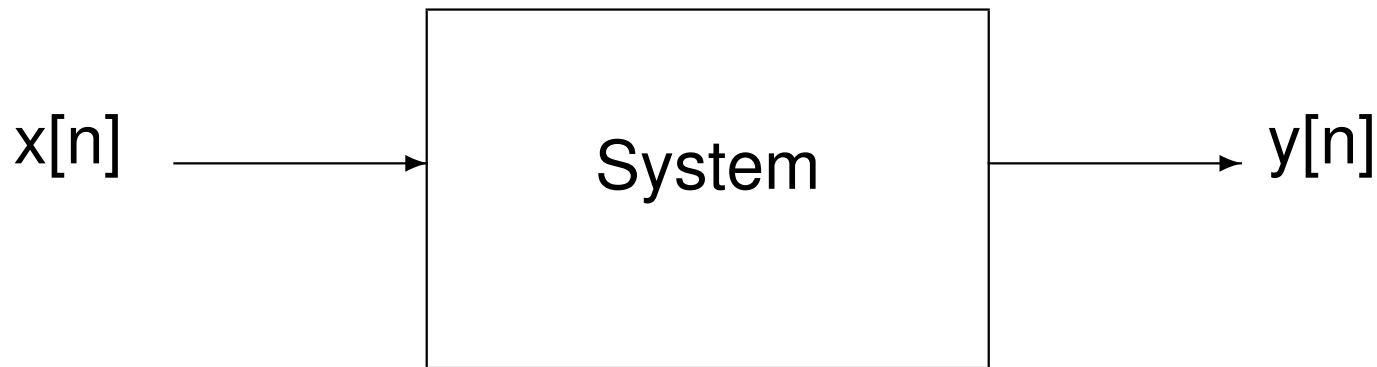


# Lecture: Introduction to Systems and FIR filters

# Systems

- ▶ A **system** is used to process an input signal  $x[n]$  and produce the output signal  $y[n]$ .
  - ▶ We focus on discrete-time signals and systems;
  - ▶ a corresponding theory exists for continuous-time signals and systems.
- ▶ Many different systems:
  - ▶ Filters: remove undesired signal components,
  - ▶ Modulators and demodulators,
  - ▶ Detectors.



## Representative Examples

- ▶ The following are examples of systems:
  - ▶ **Squarer:**  $y[n] = (x[n])^2$ ;
  - ▶ **Modulator:**  $y[n] = x[n] \cdot \cos(2\pi f_d n)$ ;
  - ▶ **Averager:**  $y[n] = \frac{1}{M} \sum_{k=0}^{M-1} x[n-k]$ ;
  - ▶ **FIR Filter:**  $y[n] = \sum_{k=0}^M b_k x[n-k]$
- ▶ In MATLAB, systems are generally modeled as functions with  $x[n]$  as the first input argument and  $y[n]$  as the output argument.
  - ▶ **Example:** first two lines of function implementing a squarer.

```
function yy = squarer(xx)
% squarer - output signal is the square of the input signal
```

# Squarer

- ▶ System relationship between input and output signals:

$$y[n] = (x[n])^2.$$

- ▶ **Example:** Input signal:  $x[n] = \{1, 2, 3, 4, 3, 2, 1\}$ 
  - ▶ **Notation:**  $x[n] = \{1, 2, 3, 4, 3, 2, 1\}$  means  
 $x[0] = 1, x[1] = 2, \dots, x[6] = 1$ ;  
 all other  $x[n] = 0$ .
- ▶ Output signal:  $y[n] = \{1, 4, 9, 16, 9, 4, 1\}$ .

## Modulator

- ▶ System relationship between input and output signals:

$$y[n] = (x[n]) \cdot \cos(2\pi f_d n);$$

where the modulator frequency  $f_d$  is a *parameter* of the system.

- ▶ **Example:**

- ▶ Input signal:  $x[n] = \{1, 2, 3, 4, 3, 2, 1\}$
- ▶ assume  $f_d = 0.5$ , i.e.,  $\cos(2\pi f_d n) = \{\dots, 1, -1, 1, -1, \dots\}$ .
- ▶ Output signal:  $y[n] = \{1, -2, 3, -4, 3, -2, 1\}$ .

## Averager

- ▶ System relationship between input and output signals:

$$\begin{aligned}
 y[n] &= \frac{1}{M} \sum_{k=0}^{M-1} x[n-k] \\
 &= \frac{1}{M} \cdot (x[n] + x[n-1] + \dots + x[n-(M-1)]) \\
 &= \sum_{k=0}^{M-1} \frac{1}{M} \cdot x[n-k].
 \end{aligned}$$

- ▶ This system computes the *sliding average* over the  $M$  most recent samples.
- ▶ **Example:** Input signal:  $x[n] = \{1, 2, 3, 4, 3, 2, 1\}$
- ▶ For computing the output signal, a table is very useful.
  - ▶ **synthetic multiplication** table.

## 3-Point Averager ( $M = 3$ )

$n$	-1	0	1	2	3	4	5	6	7	8
$x[n]$	0	1	2	3	4	3	2	1	0	0
$\frac{1}{M} \cdot x[n]$	0	$\frac{1}{3}$	$\frac{2}{3}$	1	$\frac{4}{3}$	1	$\frac{2}{3}$	$\frac{1}{3}$	0	0
$+\frac{1}{M} \cdot x[n-1]$	0	0	$\frac{1}{3}$	$\frac{2}{3}$	1	$\frac{4}{3}$	1	$\frac{2}{3}$	$\frac{1}{3}$	0
$+\frac{1}{M} \cdot x[n-2]$	0	0	0	$\frac{1}{3}$	$\frac{2}{3}$	1	$\frac{4}{3}$	1	$\frac{2}{3}$	$\frac{1}{3}$
$y[n]$	0	$\frac{1}{3}$	1	2	3	$\frac{10}{3}$	3	2	1	$\frac{1}{3}$

►  $y[n] = \left\{ \frac{1}{3}, 1, 2, 3, \frac{10}{3}, 3, 2, 1, \frac{1}{3} \right\}$

## General FIR Filter

- ▶ The M-point averager is a special case of the general **FIR filter**.
  - ▶ FIR stands for **Finite Impulse Response**; we will see what this means later.
- ▶ The system relationship between the input  $x[n]$  and the output  $y[n]$  is given by

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

- ▶  $M$  is the number of filter coefficients.
- ▶  $M - 1$  is called the **order** of the filter.



## General FIR Filter

- ▶ System relationship:

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

- ▶ The **filter coefficients**  $b_k$  determine the characteristics of the filter.
  - ▶ Much more on the relationship between the filter coefficients  $b_k$  and the characteristics of the filter later.
- ▶ Clearly, with  $b_k = \frac{1}{M}$  for  $k = 0, 1, \dots, M - 1$  we obtain the M-point averager.
- ▶ Again, computation of the output signal can be done via a synthetic multiplication table.
  - ▶ **Example:**  $x[n] = \{1, 2, 3, 4, 3, 2, 1\}$  and  $b_k = \{1, -2, 1\}$ .

## FIR Filter ( $b_k = \{1, -2, 1\}$ )

$n$	-1	0	1	2	3	4	5	6	7	8
$x[n]$	0	1	2	3	4	3	2	1	0	0
$1 \cdot x[n]$	0	1	2	3	4	3	2	1	0	0
$-2 \cdot x[n-1]$	0	0	-2	-4	-6	-8	-6	-4	-2	0
$+1 \cdot x[n-2]$	0	0	0	1	2	3	4	3	2	1
$y[n]$	0	1	0	0	0	-2	0	0	0	1

- ▶  $y[n] = \{1, 0, 0, 0, -2, 0, 0, 0, 1\}$
- ▶ Note that the output signal  $y[n]$  is longer than the input signal  $x[n]$ .
- ▶ Note, synthetic multiplication works only for short, finite-duration signal.

# Exercise

1. Find the output signal  $y[n]$  for an FIR filter

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

with filter coefficients  $b_k = \{1, -1, 2\}$  when the input signal is  $x[n] = \{1, 2, 4, 2, 4, 2, 1\}$ .

# Unit Step Sequence and Unit Step Response

- ▶ The signal with samples

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

is called the **unit-step sequence** or **unit-step signal**.

- ▶ The output of an FIR filter when the input is the unit-step signal ( $x[n] = u[n]$ ) is called the **unit-step response**  $r[n]$ .



# Unit-Step Response of the 3-Point Averager

- ▶ Input signal:  $x[n] = u[n]$ .
- ▶ Output signal:  $r[n] = \frac{1}{3} \sum_{k=0}^2 u[n - k]$ .

$n$	-1	0	1	2	3	...
$u[n]$	0	1	1	1	1	...
$\frac{1}{3}u[n]$	0	$\frac{1}{3}$	$\frac{1}{3}$	$\frac{1}{3}$	$\frac{1}{3}$	...
$+\frac{1}{3}u[n-1]$	0	0	$\frac{1}{3}$	$\frac{1}{3}$	$\frac{1}{3}$	...
$+\frac{1}{3}u[n-2]$	0	0	0	$\frac{1}{3}$	$\frac{1}{3}$	...
$r[n]$	0	$\frac{1}{3}$	$\frac{2}{3}$	1	1	...

# Unit-Impulse Sequence and Unit-Impulse Response

- ▶ The signal with samples

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

is called the **unit-impulse sequence** or **unit-impulse signal**.

- ▶ The output of an FIR filter when the input is the unit-impulse signal ( $x[n] = \delta[n]$ ) is called the **unit-impulse response**, denoted  $h[n]$ .
- ▶ Typically, we will simply call the above signals simply **impulse signal** and **impulse response**.
- ▶ We will see that the impulse-response captures all characteristics of a FIR filter.
  - ▶ This implies that impulse response is a very important concept!

# Unit-Impulse Response of a FIR Filter

- ▶ Input signal:  $x[n] = \delta[n]$ .
- ▶ Output signal:  $h[n] = \sum_{k=0}^{M-1} b_k \delta[n - k]$ .

$n$	-1	0	1	2	3	...	M
$\delta[n]$	0	1	0	0	0	...	0
$b_0 \cdot \delta[n]$	0	$b_0$	0	0	0	...	0
$+ b_1 \cdot \delta[n - 1]$	0	0	$b_1$	0	0	...	0
$+ b_2 \cdot \delta[n - 2]$	0	0	0	$b_2$	0	...	0
$\vdots$				$\vdots$			
$+ b_M \cdot \delta[n - M]$	0	0	0	0	0	...	$b_M$
$h[n]$	0	$b_0$	$b_1$	$b_2$	$b_3$	...	$b_M$



## Important Insights

- ▶ For an FIR filter, the impulse response equals the sequence of filter coefficients:

$$h[n] = \begin{cases} b_n & \text{for } n = 0, 1, \dots, M - 1 \\ 0 & \text{else.} \end{cases}$$

- ▶ Because of this relationship, the system relationship for an FIR filter can also be written as

$$\begin{aligned} y[n] &= \sum_{k=0}^{M-1} b_k x[n - k] \\ &= \sum_{k=0}^{M-1} h[k] x[n - k] \\ &= \sum_{-\infty}^{\infty} h[k] x[n - k]. \end{aligned}$$

- ▶ The operation  $y[n] = h[n] * x[n] = \sum_{-\infty}^{\infty} h[k] x[n - k]$  is called **convolution**; it is a **very, very** important operation.