

Lecture 2

Linear Time-Invariant System



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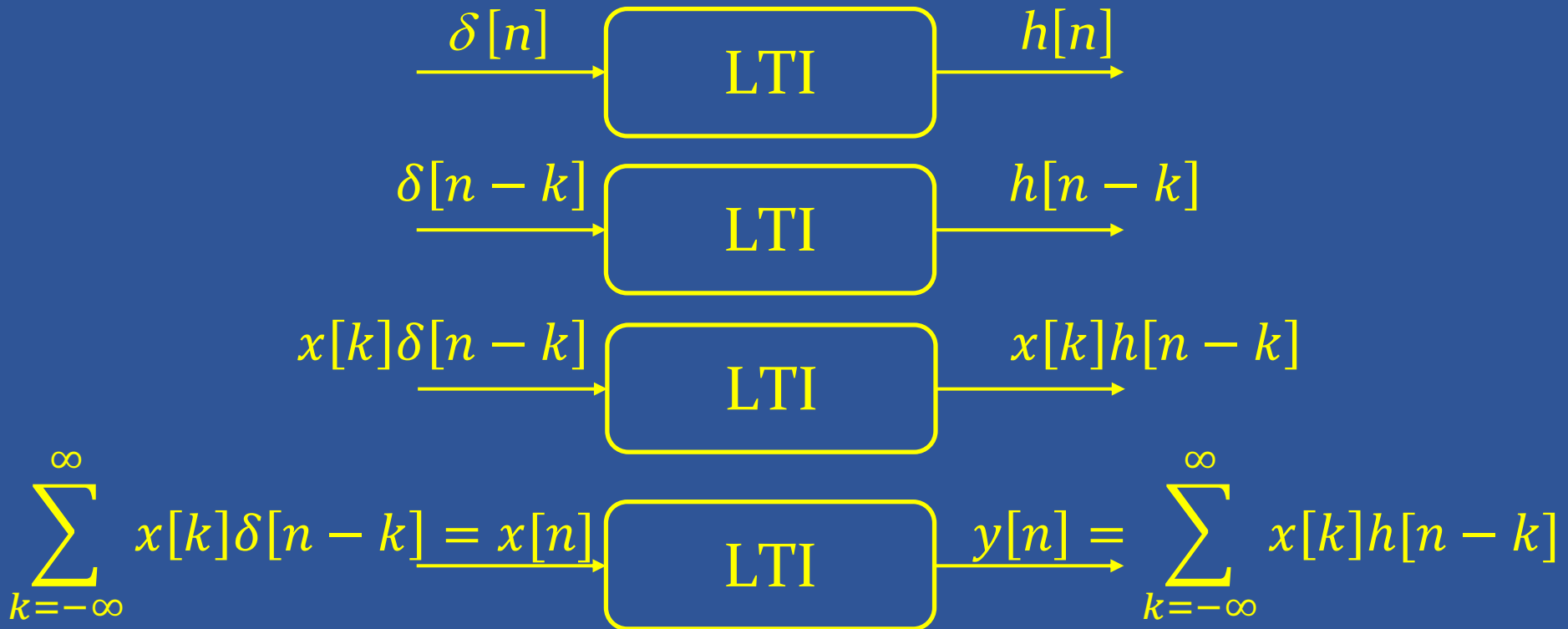


LTI Systems

- ❑ Linear Time-Invariant (LTI) System – a system satisfying the **linearity** and **time-invariant** properties
- ❑ LTI systems are mathematically **easy to analyze** and **characterize**, and consequently **easy to design**
- ❑ Highly useful signal processing algorithms have been developed utilizing this class of systems

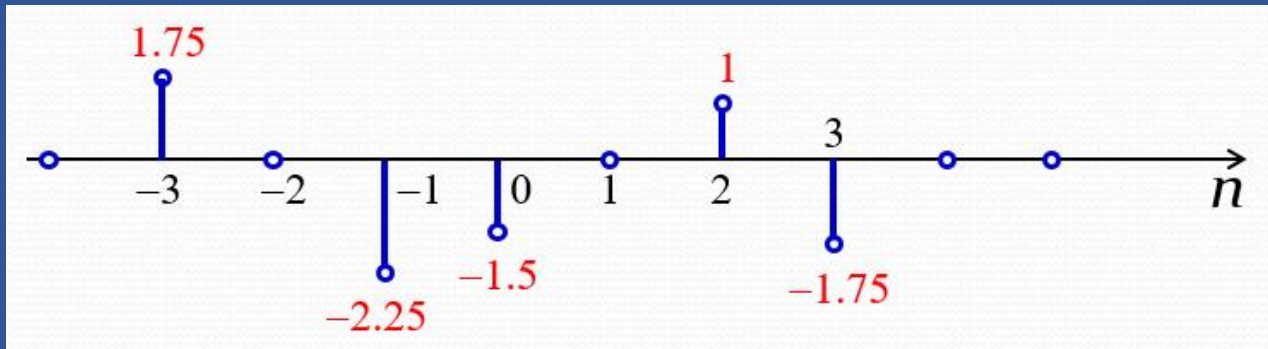


Why LTI System?



Discrete LTI Systems: Arbitrary Sequence

- An arbitrary sequence can be represented as the weighted sum of shifted unit impulses



$$x[n] = 1.75\delta[n+3] - 2.25\delta[n+1] - 1.5\delta[n] + \delta[n-2] - 1.75\delta[n-3]$$

- A general form

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

Sifting property of $\delta[n]$

Impulse Response

- The response of a system to a **unit impulse** sequence $\{\delta[n]\}$ is called **impulse response**, denoted by $\{h[n]\}$



Impulse Response

□ The **impulse response** of a system

$$y[n] = a_1x[n] + a_2x[n-1] + a_3x[n-2] + a_4x[n-3]$$

is obtained by setting $x[n] = \delta[n]$, resulting in

$$h[n] = a_1\delta[n] + a_2\delta[n-1] + a_3\delta[n-2] + a_4\delta[n-3]$$

□ The impulse response is a finite length sequence of length 4, given by

$$\{h[n]\} = \{a_1, a_2, a_3, a_4\}$$



Example

□ The impulse response of $y[n] = \sum_{k=-\infty}^n x[k]$

By setting $x[n] = \delta[n]$, we have

$$h(n) = \sum_{k=-\infty}^n \delta[k]$$

which is precisely the **unit step sequence**



Example

- The impulse response of factor-of-2 interpolator

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

is obtained by setting $x_u[n] = \delta[n]$, resulting in

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

- The impulse response is thus a finite length sequence of length 3 given by

$$h[n] = \{0.5, 1, 0.5\}$$



Time-Domain Characterization

- ❑ Input-output relationship – An LTI discrete system is **completely** characterized by its impulse response
- ❑ In other words, **knowing the impulse response one can compute the output of the system for an arbitrary input**

“DNA of LTI”



Output of LTI Systems

□ Let $h[n]$ denote the impulse response of a discrete **LTI** system, compute the output $y[n]$ for the input:

$$x[n] = 0.5\delta[n + 2] + 1.5\delta[n - 1] - \delta[n - 2] + 0.75\delta[n - 5]$$



Output of LTI Systems

□ Since the system is **time-invariant**, we have

Input		Output
$\delta[n + 2]$	\rightarrow	
$\delta[n - 1]$	\rightarrow	
$\delta[n - 2]$	\rightarrow	
$\delta[n - 5]$	\rightarrow	

Then $x[n - n_0] \rightarrow y[n - n_0]$



Output of LTI Systems

□ Likewise, since the system is **linear**, we have

Input	Output
$0.5\delta[n + 2]$	\rightarrow
$1.5\delta[n - 1]$	\rightarrow
$\delta[n - 2]$	\rightarrow
$0.75\delta[n - 5]$	\rightarrow

□ Hence, **because of the linear property**, we get

$$y[n] = 0.5h[n + 2] + 1.5h[n - 1] - h[n - 2] + 0.75h[n - 5]$$



Output of LTI Systems

- Recall, an arbitrary input $x[n]$ can be expressed as a **linear combination** of shifted **unit impulses**

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

- The response of a **LTI** system to $x[k]\delta[n-k]$ will be $x[k]h[n-k]$

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



Impulse Response: DNA of LTI

- The impulse response $h[n]$ completely characterizes an LTI system

“DNA of LTI”



$$x[n] = \sum_{k=-\infty}^{\infty} x[k]\delta[n - k] \rightarrow \text{LTI} \rightarrow y[n] = \sum_{k=-\infty}^{\infty} x[k]h[n - k]$$



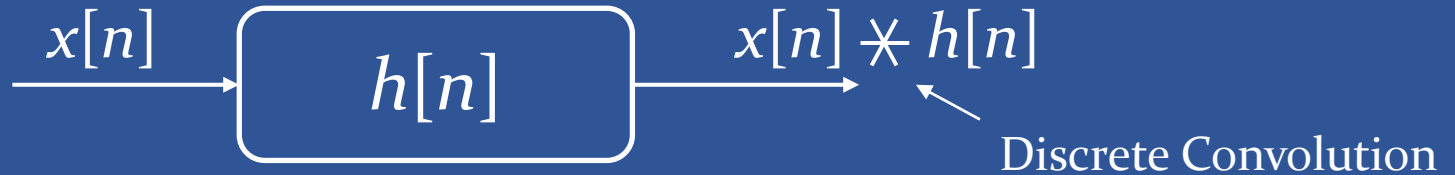
Impulse Response: DNA of LTI

□ Mathematically,

$$\begin{aligned}y[n] &= \text{LTI}\{x[n]\} = \text{LTI}\left\{\sum_{k=-\infty}^{\infty} x[k]\delta[n-k]\right\} \\&= \sum_{k=-\infty}^{\infty} x[k]\text{LTI}\{\delta[n-k]\} \\&= \sum_{k=-\infty}^{\infty} x[k]h[n-k]\end{aligned}$$



The Discrete Convolution



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

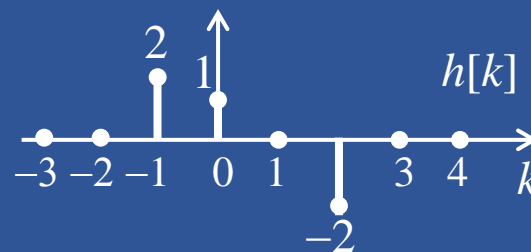
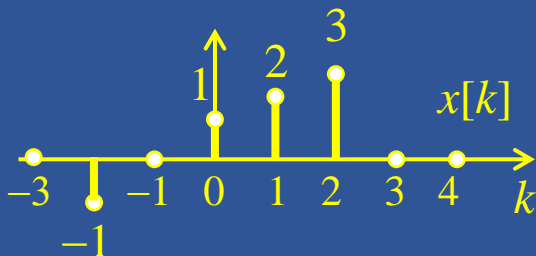
The above summation is defined to be the convolution of the sequences $x[n]$ and $h[n]$ and represented compactly as

$$y[n] = x[n] * h[n]$$



Discrete Convolution: An Example

Compute the convolution of $x[n]$ and $h[n]$

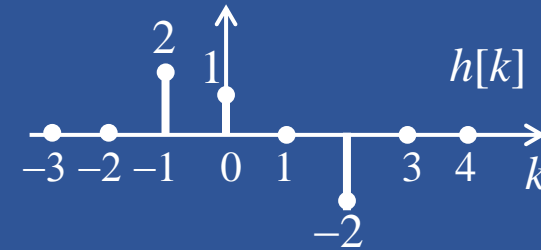
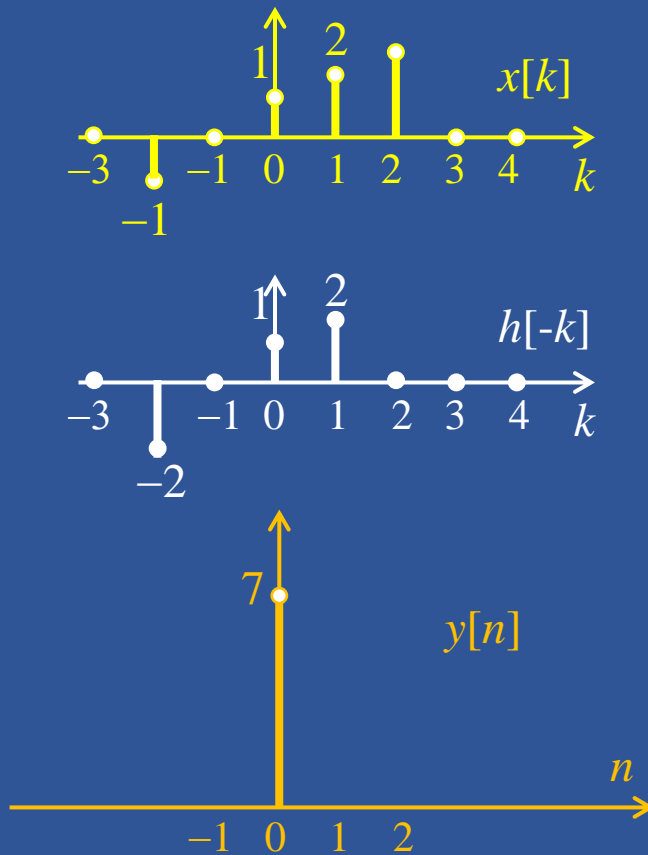


$$y[n] = \sum_{k=-\infty}^{\infty} x[k]h[n-k]$$

$$y[0] = \sum_{k=-\infty}^{\infty} x[k]h[-k]$$



Discrete Convolution: An Example

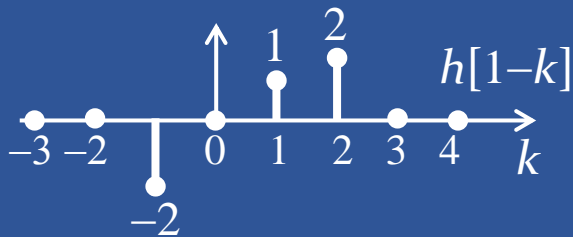
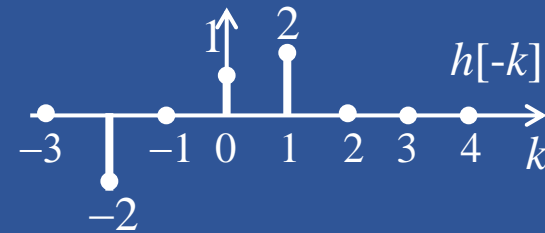
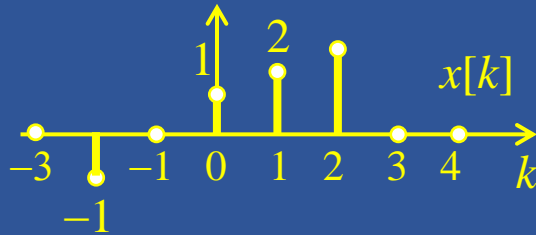
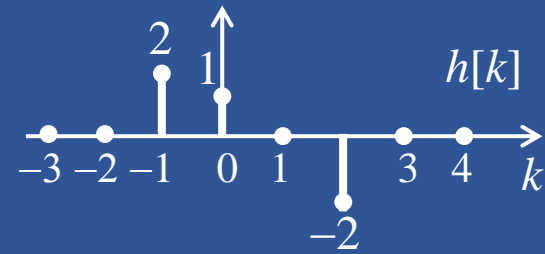


$$y[n] = \sum_{k=-\infty}^{\infty} x[k]h[n-k]$$

$$y[0] = \sum_{k=-\infty}^{\infty} x[k]h[-k]$$

$$\begin{aligned} &= x[-2]h[2] + x[-1]h[1] + x[0]h[0] \\ &\quad + x[1]h[-1] + x[2]h[-2] \\ &= -1 \times (-2) + 1 \times 1 + 2 \times 2 = 7 \end{aligned}$$



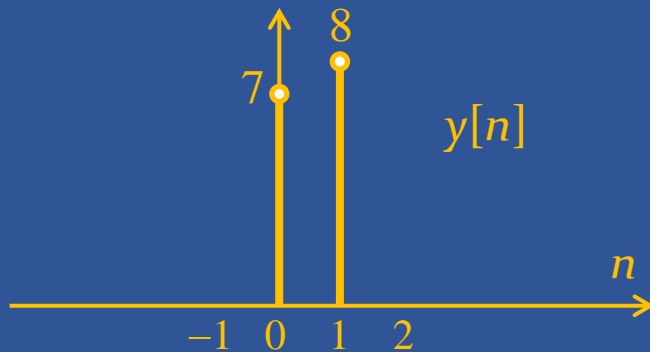


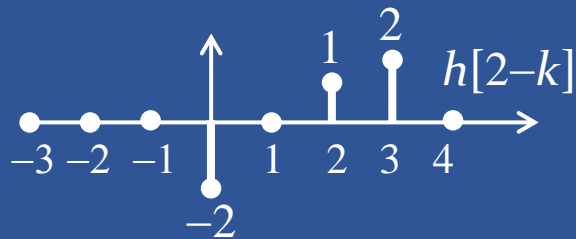
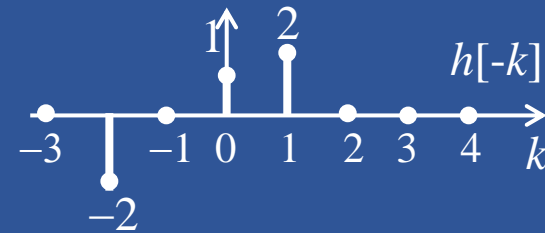
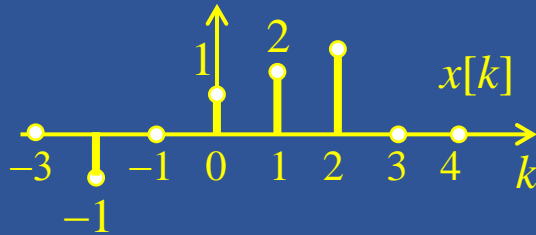
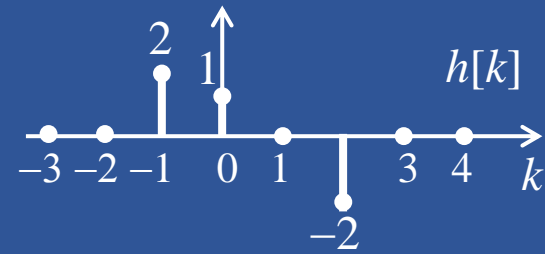
$$y[n] = \sum_{k=-\infty}^{\infty} x[k]h[n-k]$$

$$y[1] = \sum_{k=-\infty}^{\infty} x[k]h[1-k]$$

$$= x[-2]h[3] + x[-1]h[2] + x[0]h[1] \\ + x[1]h[0] + x[2]h[-1]$$

$$= 2 \times 1 + 3 \times 2 = 8$$



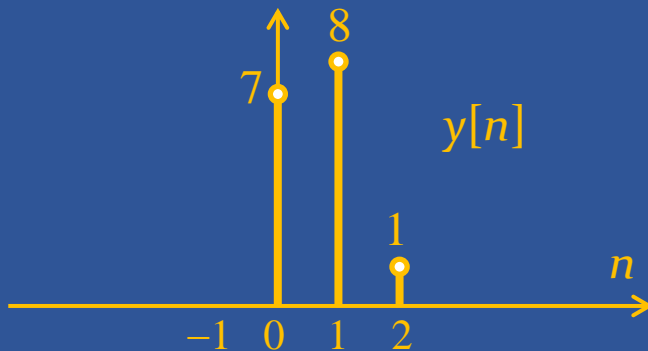


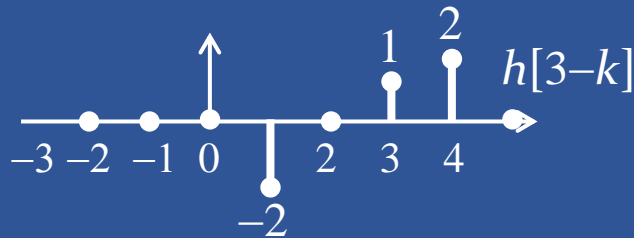
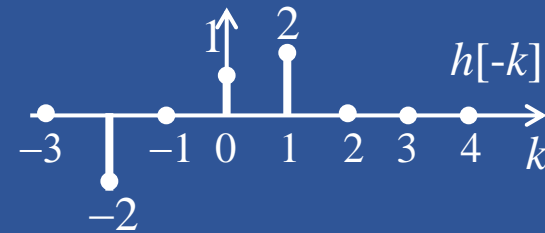
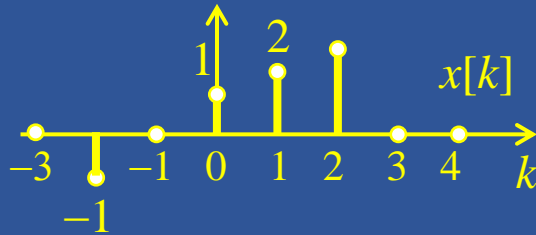
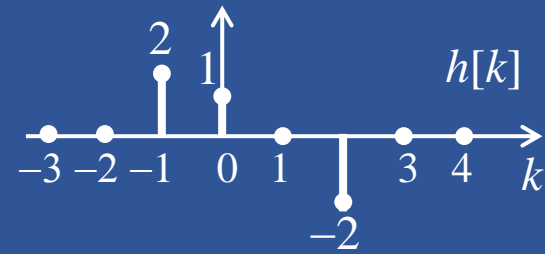
$$y[n] = \sum_{k=-\infty}^{\infty} x[k]h[n-k]$$

$$y[2] = \sum_{k=-\infty}^{\infty} x[k]h[2-k]$$

$$= x[-2]h[4] + x[-1]h[3] + x[0]h[2] \\ + x[1]h[1] + x[2]h[0]$$

$$= 1 \times (-2) + 3 \times 1 = 1$$



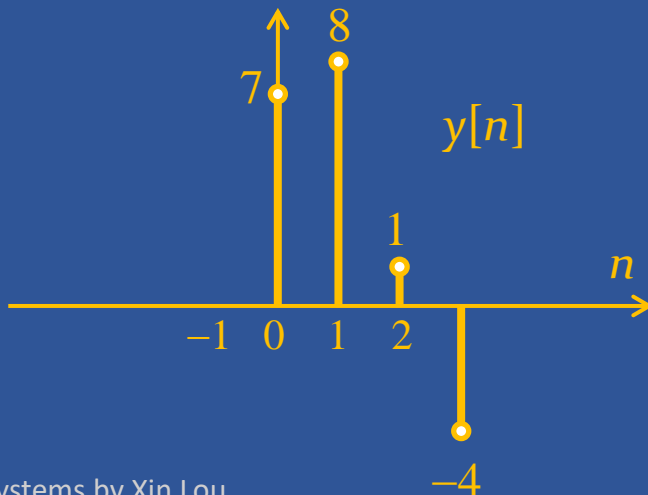


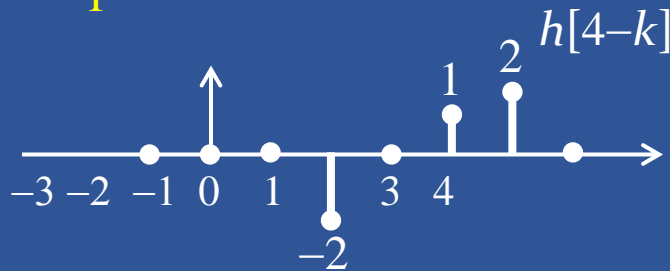
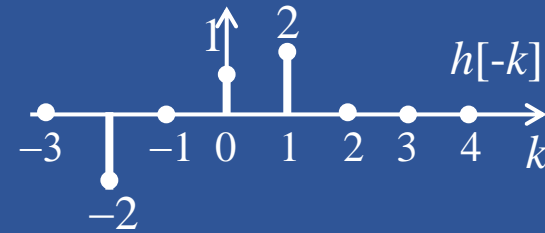
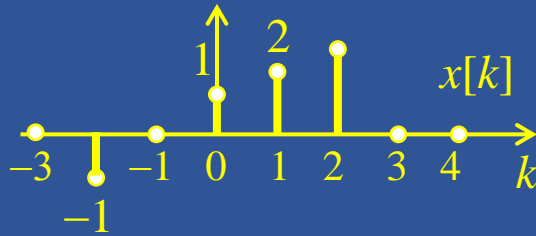
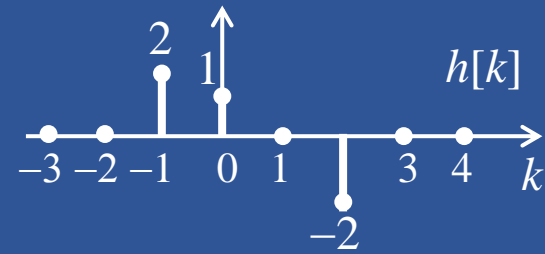
$$y[n] = \sum_{k=-\infty}^{\infty} x[k]h[n-k]$$

$$y[3] = \sum_{k=-\infty}^{\infty} x[k]h[3-k]$$

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

$$= 2 \times (-2) = -4$$



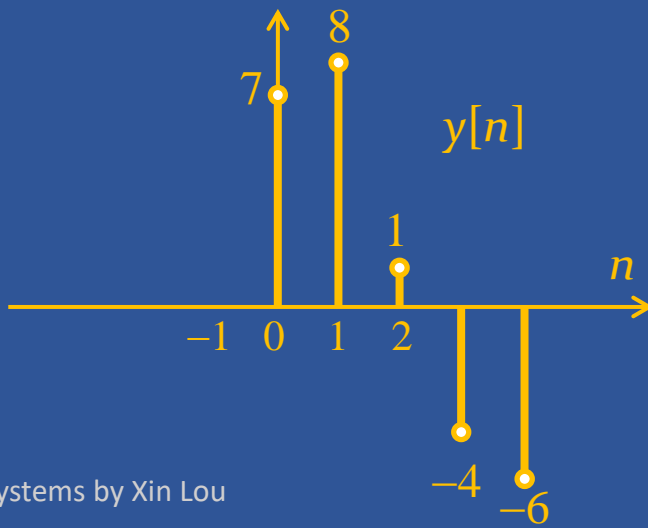


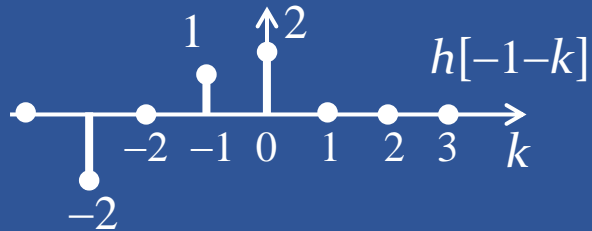
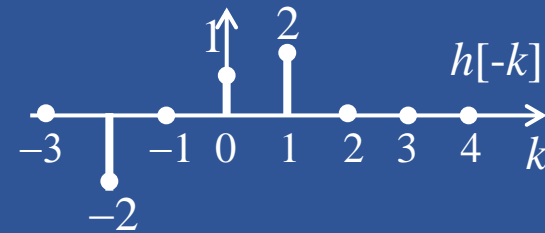
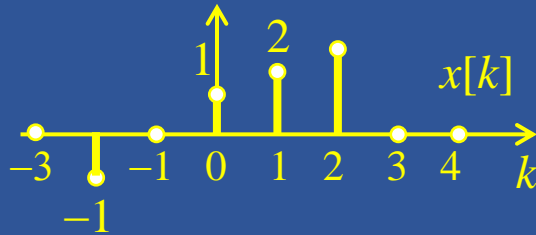
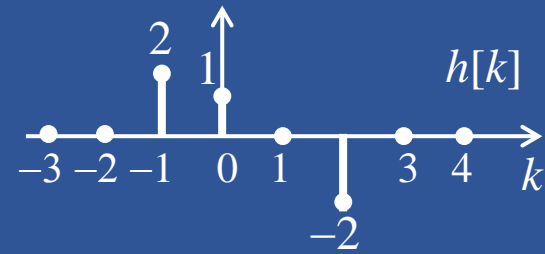
$$y[n] = \sum_{k=-\infty}^{\infty} x[k]h[n-k]$$

$$y[4] = \sum_{k=-\infty}^{\infty} x[k]h[4-k]$$

$$= x[-2]h[6] + x[-1]h[5] + x[0]h[4] \\ + x[1]h[3] + x[2]h[2]$$

$$= 3 \times (-2) = -6$$

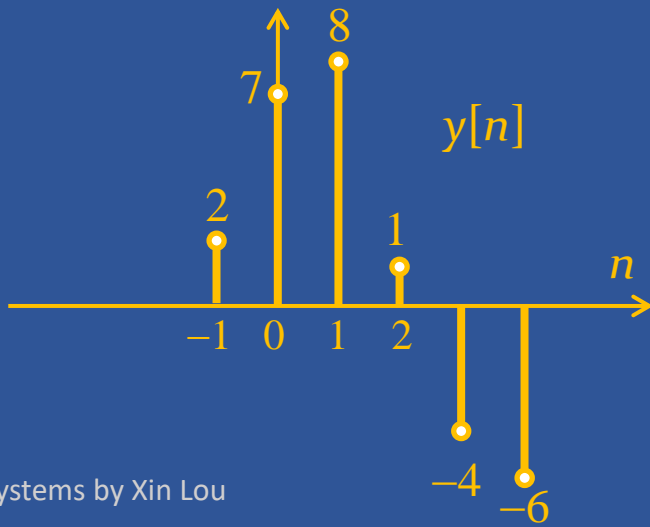


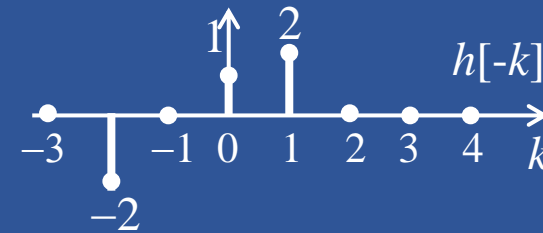
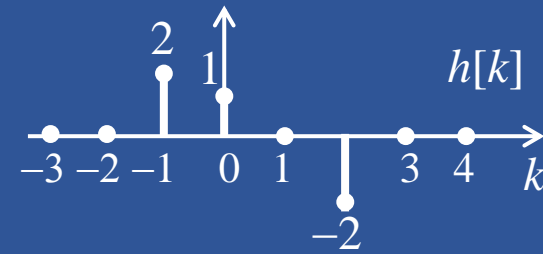


$$y[n] = \sum_{k=-\infty}^{\infty} x[k]h[n-k]$$

$$y[-1] = \sum_{k=-\infty}^{\infty} x[k]h[-1-k]$$

$$\begin{aligned} &= x[-2]h[1] + x[-1]h[0] + x[0]h[-1] \\ &\quad + x[1]h[-2] + x[2]h[-3] \\ &= 1 \times 2 = 2 \end{aligned}$$

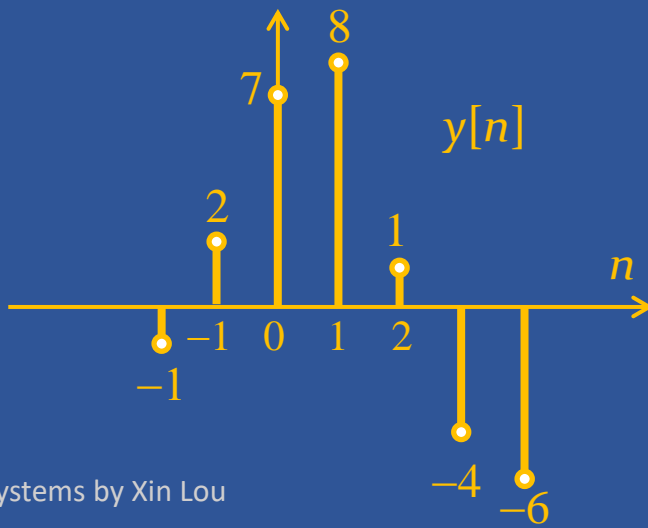
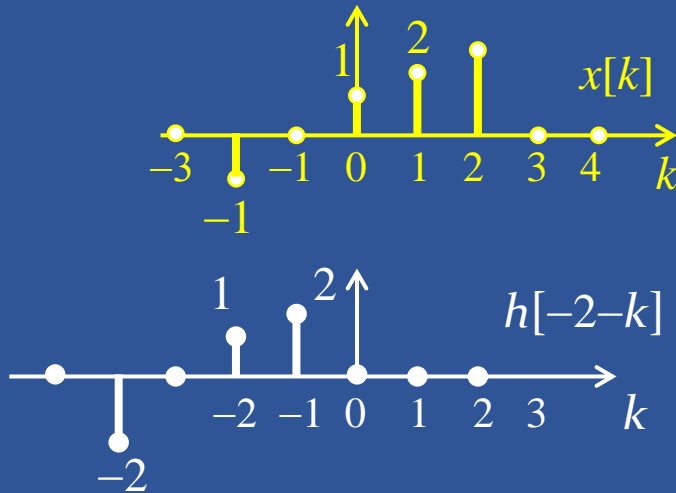


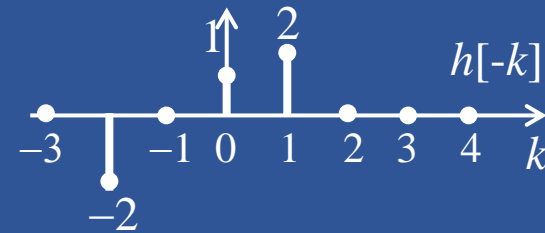
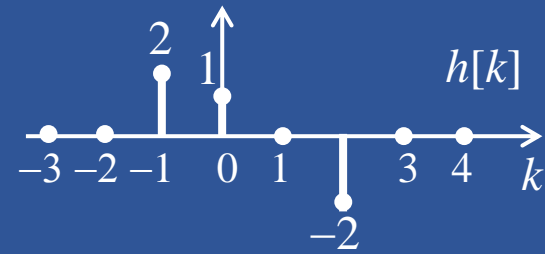


$$y[n] = \sum_{k=-\infty}^{\infty} x[k]h[n-k]$$

$$y[-2] = \sum_{k=-\infty}^{\infty} x[k]h[-2-k]$$

$$\begin{aligned} &= x[-2]h[0] + x[-1]h[-1] + x[0]h[-2] \\ &\quad + x[1]h[-3] + x[2]h[-4] \\ &= -1 \times 1 = -1 \end{aligned}$$

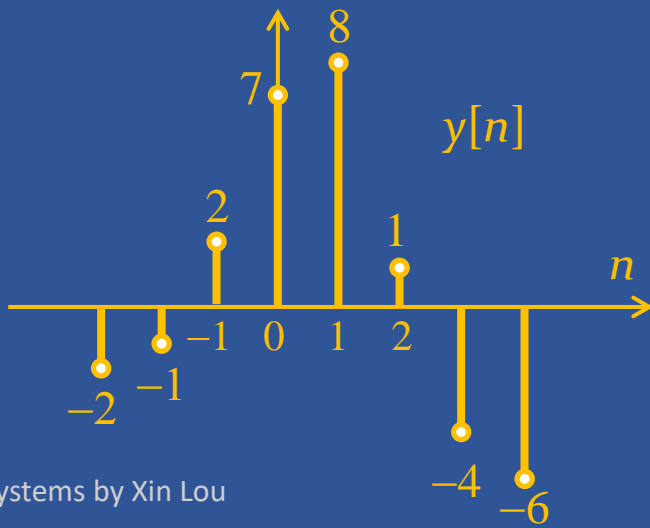
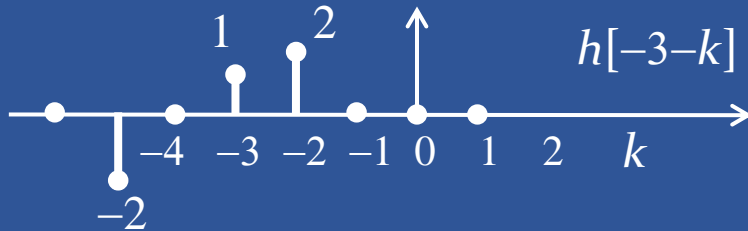
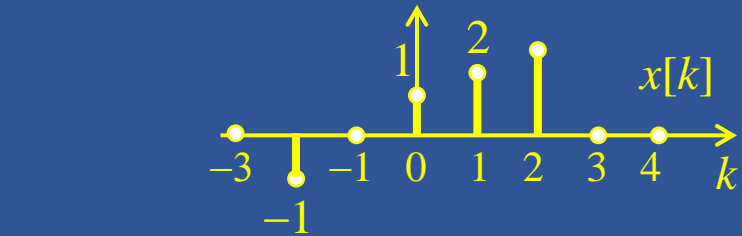




$$y[n] = \sum_{k=-\infty}^{\infty} x[k]h[n-k]$$

$$y[-3] = \sum_{k=-\infty}^{\infty} x[k]h[-3-k]$$

$$\begin{aligned} &= x[-2]h[-1] + x[-1]h[-2] + x[0]h[-3] \\ &\quad + x[1]h[-4] + x[2]h[-5] \\ &= -1 \times 2 = -2 \end{aligned}$$



Computation of Discrete Convolution

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

- ❑ Fold $h[k]$ with respect to the origin to obtain $h[-k]$
- ❑ Shift to right (if $n \geq 0$) or to the left (if $n < 0$) by $|n|$ samples
- ❑ Compute the products of the corresponding samples of sequences $h[n-k]$ and $x[k]$
- ❑ Sum the all products to obtain $y[n]$
- ❑ **Fold, shift, product, and sum**



Computation of Convolution

- The sum of indices of each sample product inside the convolution sum is equal to the index of the sample being computed

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

$$\begin{aligned} y[-3] &= \sum_{k=-\infty}^{\infty} x[k]h[-3-k] \\ &= x[-2]h[-1] + x[-1]h[-2] + x[0]h[-3] + x[1]h[-4] + x[2]h[-5] \end{aligned}$$

- If the lengths of the two sequences are M and N , then the sequence generated by the convolution is of length $M+N-1$



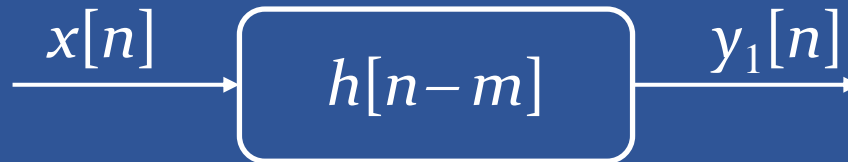
Example

□ If



$$\text{i.e., } y[n] = x[n] * h[n] = \sum_{k=-\infty}^{\infty} x[k]h[n-k]$$

□ How about



$$y_1[n] = x[n] * h[n-m] = \sum_{k=-\infty}^{\infty} x[k]h[n-m-k]$$

$$= y[n-m]$$



Properties of Convolution

- Sequence shifting is equivalent to convolve with a shifted impulse

$$x[n] * \delta[n-d] = x[n-d]$$



Continuous LTI Systems: Arbitrary Signals

□ In the discrete case, we have

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

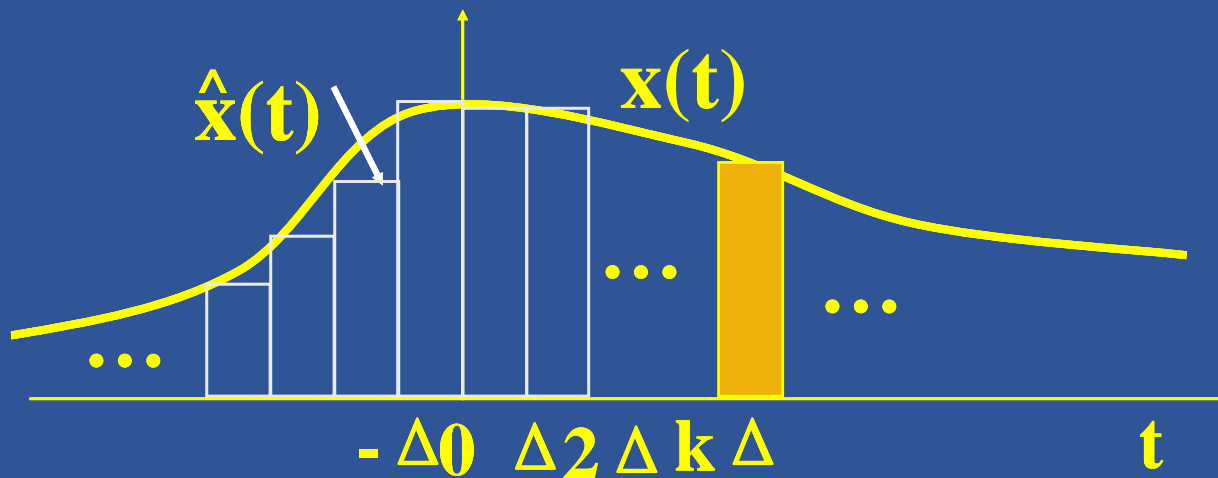
□ In the continuous case, we also have

$$x(t) = \int_{-\infty}^{\infty} x(\tau)\delta(t-\tau)d\tau$$



Continuous LTI Systems: Arbitrary Signals

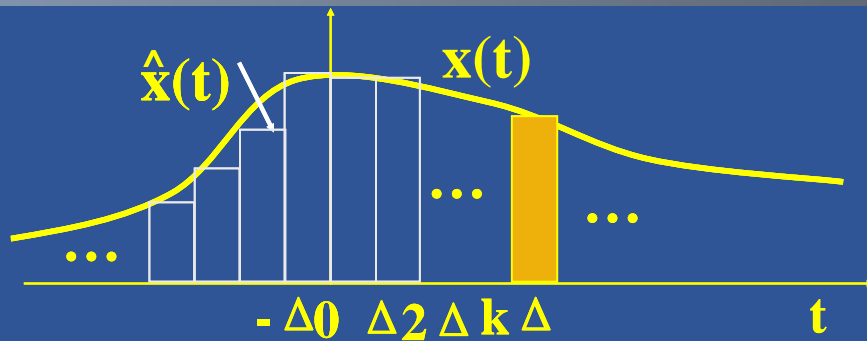
- Consider a pulse or “staircase” approximation, $\hat{x}(t)$ to a continuous signal $x(t)$.



$$\delta_{\Delta}(t) = \begin{cases} \frac{1}{\Delta}, & 0 \leq t \leq \Delta \\ 0, & \text{otherwise} \end{cases}$$



Continuous LTI Systems: Arbitrary Signals



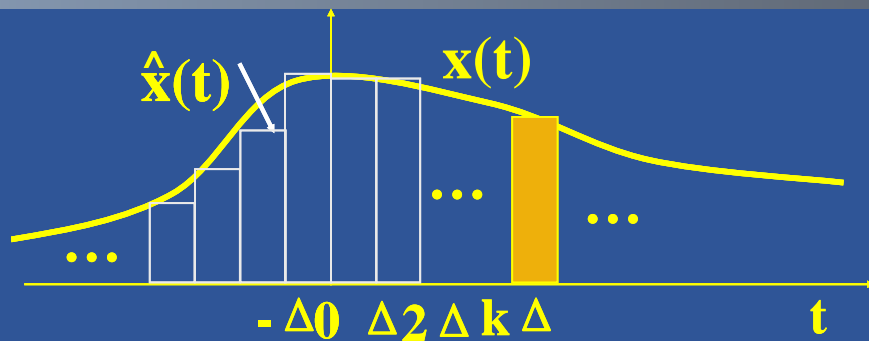
$$\delta_{\Delta}(t) = \begin{cases} \frac{1}{\Delta}, & 0 \leq t \leq \Delta \\ 0, & \text{otherwise} \end{cases}$$

□ Then , $\Delta \delta_{\Delta}(t) = 1$, the shade pulse is $x(k\Delta)\delta_{\Delta}(t - k\Delta)\Delta$

□ If we sum up all the shapes, we get

$$\hat{x}(t) = \sum_{k=-\infty}^{\infty} x(k\Delta)\delta_{\Delta}(t - k\Delta)\Delta$$

Continuous LTI Systems: Arbitrary Signals



$$\hat{x}(t) = \sum_{k=-\infty}^{\infty} x(k\Delta) \delta_{\Delta}(t - k\Delta) \Delta$$

As $\Delta \rightarrow 0$, Δ becomes $d\tau$

$$\hat{x}(t) \rightarrow x(t) \quad \sum_{k=-\infty}^{\infty} \rightarrow \int_{-\infty}^{\infty} \quad k\Delta \rightarrow \tau$$

$$x(k\Delta) \rightarrow x(\tau) \quad \delta_{\Delta}(t - k\Delta) \rightarrow \delta(t - \tau)$$

$$\sum_{k=-\infty}^{\infty} x(k\Delta) \delta_{\Delta}(t - k\Delta) \Delta \rightarrow \int_{-\infty}^{\infty} x(\tau) \delta(t - \tau) d\tau$$

Sifting property of $\delta(t)$



Example

$$x(t) = \int_{-\infty}^{\infty} x(\tau) \delta(t - \tau) d\tau$$

□ Let $x(t)=u(t)$,

$$u(t) = \int_{-\infty}^{\infty} u(\tau) \delta(t - \tau) d\tau$$

$$= \int_0^{\infty} \delta(t - \tau) d\tau$$



The Convolution Integral

□ Let $h(t)$ be the response of LTI to $\delta(t)$

$$\delta(t) \xrightarrow{\text{LTI}} h(t) \quad \text{and} \quad x(t) = \int_{-\infty}^{\infty} x(\tau) \delta(t - \tau) d\tau$$

$$\text{Q: } x(t) \xrightarrow{\text{LTI}} y(t) = ?$$

$$\text{A: Time-Invariant property} \longrightarrow \delta(t - \tau) \xrightarrow{\text{LTI}} h(t - \tau)$$

$$x(t) \xrightarrow{\text{LTI}} y(t) = \int_{-\infty}^{\infty} x(\tau) h(t - \tau) d\tau$$

$$y(t) = x(t) * h(t) = \int_{-\infty}^{\infty} x(\tau) h(t - \tau) d\tau$$

Convolution integral



Computation of Convolution Integral

- ❑ Change time variables $x(t) \rightarrow x(\tau)$ $h(t) \rightarrow h(\tau)$
and reverse $h(\tau) \rightarrow h(-\tau)$
- ❑ Shift $h(-\tau) \rightarrow h(t - \tau)$, if $t \geq 0$ toward right, otherwise toward left
- ❑ Multiply $x(\tau) \cdot h(t - \tau)$
- ❑ Integral $\int_{-\infty}^{\infty} x(\tau) \cdot h(t - \tau) d\tau$



Example

$$x(t) = \int_{-\infty}^{\infty} x(\tau) \delta(t - \tau) d\tau = x(t) * \delta(t)$$

Q: $x(t) * \delta(t - 1) = ?$

Q: $x(t) * \delta(t - 2) = ?$

A: $x(t) * \delta(t - 1)$

A: $x(t - 2)$

$$= \int_{-\infty}^{\infty} x(\tau) \delta(t - 1 - \tau) d\tau$$

$$= \int_{-\infty}^{\infty} x(\tau) \delta(t - (\tau + 1)) d\tau$$

$$= \int_{-\infty}^{\infty} x(\tau' - 1) \delta(t - \tau') d\tau'$$

$$= x(t - 1) * \delta(t)$$



Example

□ Q: $x(t) = e^{-at}u(t)$, $h(t) = u(t)$, $a > 0$
 $x(t) * h(t) = ?$

□ A: $y(t) = \int_{-\infty}^{\infty} e^{-a\tau} u(\tau) \cdot u(t - \tau) d\tau$

For $t < 0$ $x(\tau) \cdot h(t - \tau) = 0 \implies y(t) = 0$

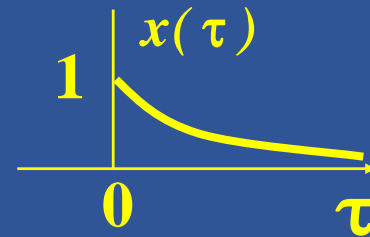
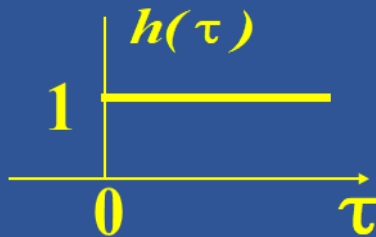
For $t \geq 0$ $y(t) = \int_0^t e^{-a\tau} d\tau$
$$= \frac{-1}{a} e^{-a\tau} \Big|_0^t = \frac{1}{a} (1 - e^{-at})$$



A Graphical Solution

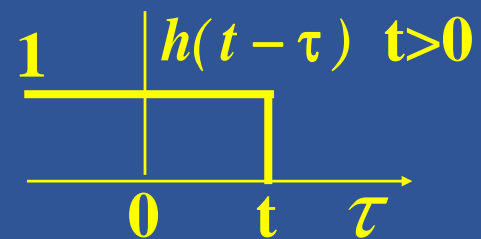
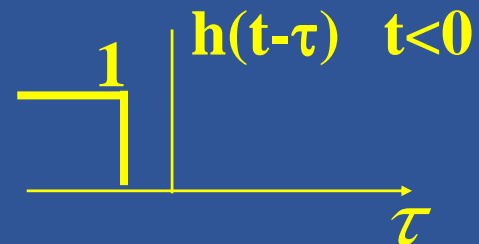
□ Q: $x(t) = e^{-at}u(t)$, $h(t) = u(t)$, $a > 0$

$$x(t) * h(t) = ?$$



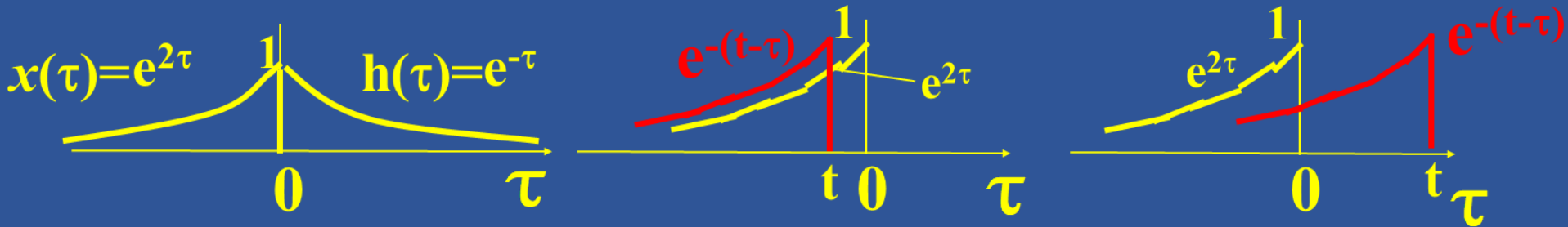
For $t \geq 0$ $y(t) = \int_0^t e^{-a\tau} d\tau$

$$= \frac{-1}{a} e^{-a\tau} \Big|_0^t = \frac{1}{a} (1 - e^{-at})$$



Example

□ Determine $y(t) = e^{2t}u(-t) * e^{-t}u(t)$



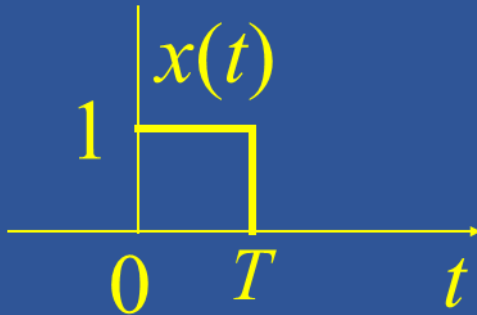
$$\begin{aligned} \square \text{ A: } y(t) &= \int_{-\infty}^{\infty} e^{2\tau}u(-\tau) \cdot e^{-(t-\tau)}u(t-\tau)d\tau \\ &= \int_{-\infty}^t e^{2\tau}e^{-(t-\tau)}d\tau + \int_{-\infty}^0 e^{-\tau}e^{-(t-\tau)}d\tau \\ &\quad (t < 0) \qquad \qquad \qquad (t > 0) \end{aligned}$$

$$= \frac{1}{3}e^{2t}u(-t) + \frac{1}{3}e^{-t}u(t)$$

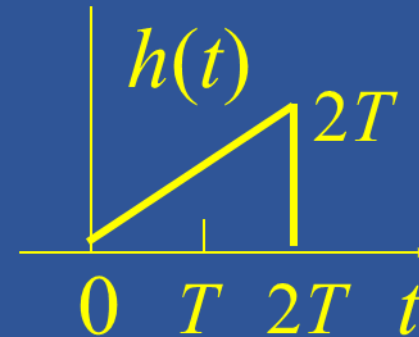


Example

$$x(t) = \begin{cases} 1, & 0 < t < T \\ 0, & \text{otherwise} \end{cases}$$



$$h(t) = \begin{cases} t, & 0 < t < 2T \\ 0, & \text{otherwise} \end{cases}$$



Determine $y(t) = x(t) * h(t)$

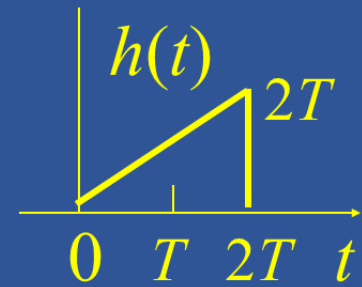
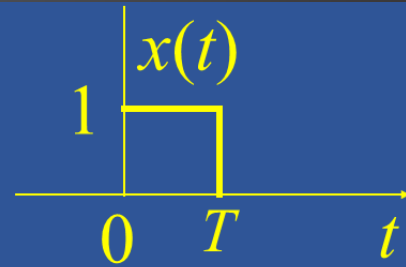


Example

1) For $t < 0$ and $t > 3T$ $y(t) = 0$

2) For $0 < t < T$

$$y(t) = \int_0^t x(\tau)h(t-\tau)d\tau = \int_0^t (t-\tau)d\tau$$
$$= t^2 - \frac{1}{2}t^2 = \frac{1}{2}t^2$$

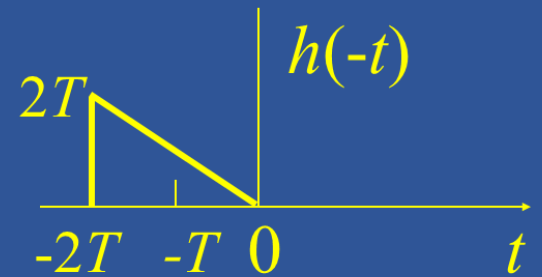


3) For $T < t < 2T$

$$y(t) = \int_0^T (t-\tau)d\tau = Tt - \frac{1}{2}T^2$$

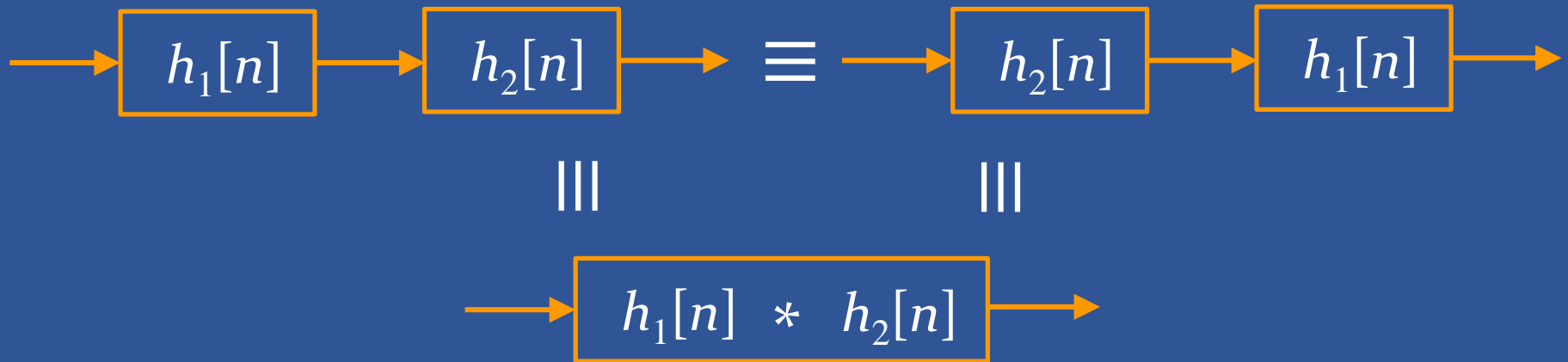
4) $2T < t < 3T$

$$y(t) = \int_{t-2T}^T (t-\tau)d\tau = -\frac{1}{2}t^2 + Tt + \frac{3}{2}T^2$$



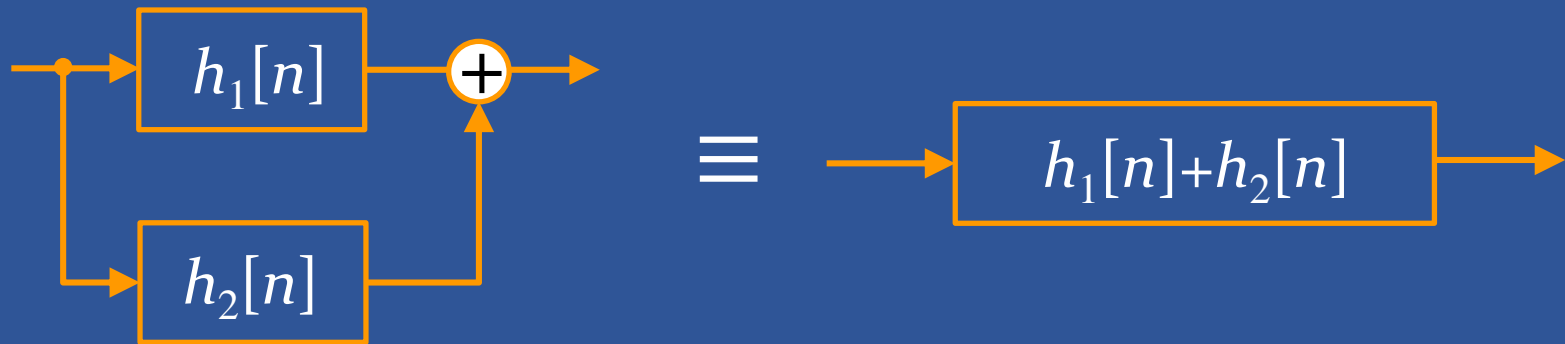
Simple Interconnection Schemes

□ Cascade Connection

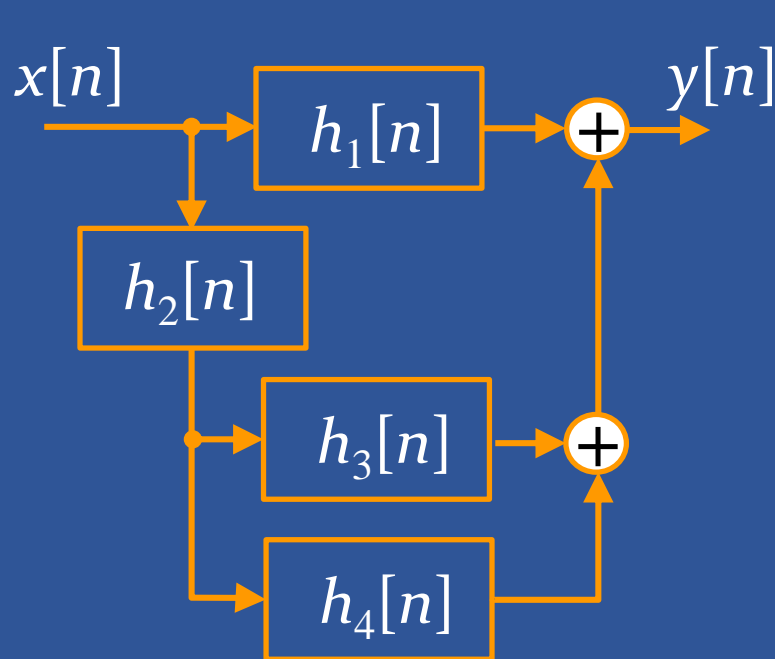


Simple Interconnection Schemes

□ Parallel Connection



Analysis of Cascade and Parallel Connections



$$y[n] = x[n] * h[n]$$

$$h[n] = h_1[n] + h_2[n] * (h_3[n] + h_4[n])$$

$$h_1[n] = \delta[n] + \frac{1}{2}\delta[n-1]$$

$$h_2[n] = \frac{1}{2}\delta[n] - \frac{1}{4}\delta[n-1]$$

$$h_3[n] = 2\delta[n]$$

$$h_4[n] = -2\left(\frac{1}{2}\right)^n \mu[n]$$

$$\rightarrow h[n] = \delta[n]$$



Properties of Convolution

□ The commutative property

$$x[n] * h[n] = h[n] * x[n]$$

$$x(t) * h(t) = h(t) * x(t)$$

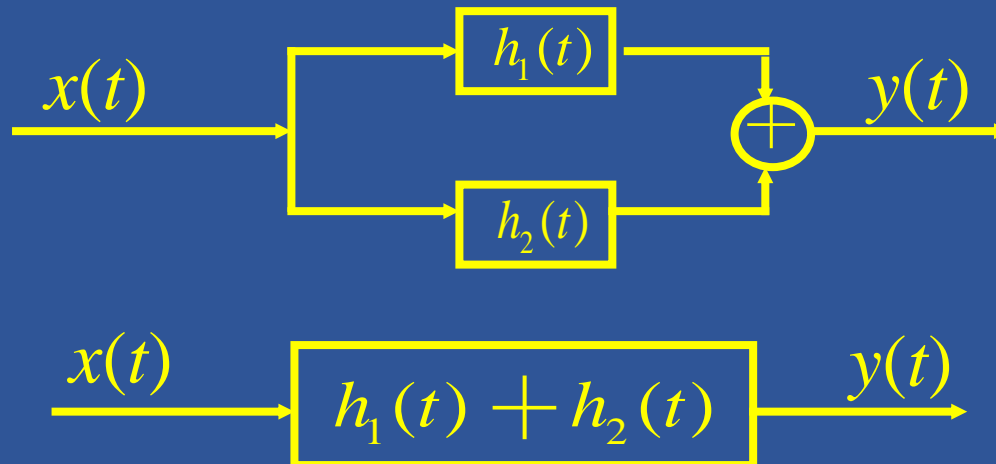


Properties of Convolution

□ The distributive property

$$x[n] * (h_1[n] + h_2[n]) = x[n] * h_1[n] + x[n] * h_2[n]$$

$$x(t) * [h_1(t) + h_2(t)] = x(t) * h_1(t) + x(t) * h_2(t)$$



Properties of Convolution

□ The associative property

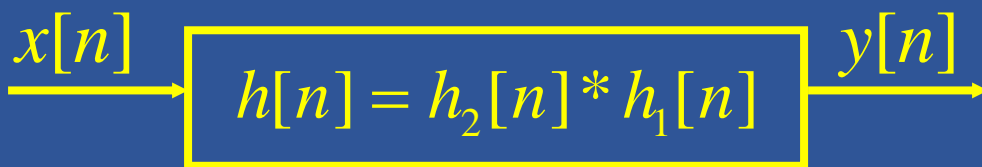
$$x[n] * (h_1[n] * h_2[n]) = (x[n] * h_1[n]) * h_2[n]$$

$$y[n] = x[n] * h_1[n] * h_2[n]$$



(a)

(b)



(c)



(d)



Properties of Convolution

- LTI systems with and without memory

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

$$y(t) = x(t) * h(t) = \int_{-\infty}^{\infty} x(\tau) h(t-\tau) d\tau$$

- For discrete system without memory:

$$h[n]=0 \text{ for } n \neq 0 \quad h[n] = k\delta[n] \quad y[n] = kx[n] \quad \text{Why?}$$

- For continuous system without memory:

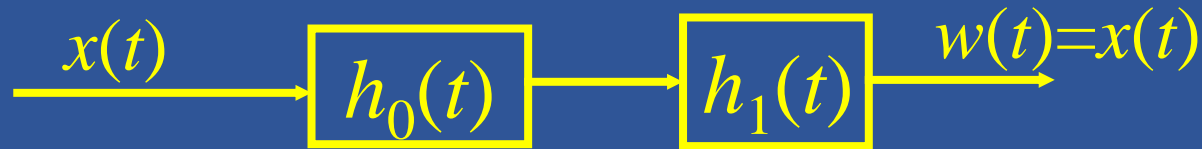
$$h(t)=0 \text{ for } t \neq 0 \quad h(t) = k\delta(t) \quad y(t) = kx(t)$$



Properties of Convolution

□ Invertibility of LTI system

➤ If $h_0(t) * h_1(t) = \delta(t)$ then the system with $h_1(t)$ is the inverse of the system with $h_0(t)$



➤ Similarly, if $h_0[n] * h_1[n] = \delta[n]$ then system of $h_1[n]$ is the inverse system of $h_0[n]$



Example

□ Consider $h_0[n] = u[n]$, determine $h_1[n]$

$$\because h_0[n] * h_1[n] = u[n] * h_1[n] \stackrel{\text{hold}}{=} \delta[n]$$

$$\delta[n] = u[n] - u[n-1] = u[n] * (\delta[n] - \delta[n-1])$$

$$\therefore h_1[n] = \delta[n] - \delta[n-1]$$



Properties of Convolution

□ Causality for LTI systems

➤ If $h[n]=0$, for $n<0$ or $h(t)=0$, for $t<0$, then the system is **causal**

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

$$y[n] = \sum_{k=-\infty}^n x[k]h[n-k] \stackrel{or}{=} \sum_{k=0}^{\infty} h[k]x[n-k]$$

$$y(t) = \int_{-\infty}^t x(\tau)h(t-\tau) \stackrel{or}{=} \int_0^{\infty} h(\tau)x(t-\tau)d\tau$$



Example

□ Causal LTI system

➤ Accumulator: $y[n] = \sum_{l=-\infty}^n x[l]$

$h[n] = \mu[n]$, a causal impulse response sequence

□ Non causal LTI system

➤ Factor-of-2 interpolator

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

$$h[n] = \{0.5, 1, 0.5\}$$

↑

a non causal impulse response sequence



Properties of Convolution

□ Stability for LTI systems

- A discrete LTI system is stable iff $h[n]$ is absolutely summable
- A continuous LTI system is stable iff $h(t)$ is absolutely integrable

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

absolutely summable

$$\int_{-\infty}^{\infty} |h(\tau)| d\tau < \infty$$

absolutely integrable



Stability of LTI Systems

□ Proof: “if” (Sufficient condition)

$$\begin{aligned} |y[n]| &= \left| \sum_{k=-\infty}^{\infty} h[k]x[n-k] \right| \leq \sum_{k=-\infty}^{\infty} |h[k]x[n-k]| \\ &= \sum_{k=-\infty}^{\infty} |h[k]| |x[n-k]| \leq B_x \\ &\leq B_x \sum_{k=-\infty}^{\infty} |h[k]| < \infty \end{aligned}$$



Stability of LTI Systems

□ Proof: “only if” (Necessary condition)

suppose $\sum_{k=-\infty}^{\infty} |h(k)| = \infty$, show that there exists

bounded $x[n]$ that gives unbounded $y[n]$

□ Let $x[n] = \frac{h[-n]}{|h[-n]|} = \text{sign}\{h[-n]\}$ Assume $h[n]$ is real sequence

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

$\text{sign}\{h[-n]\} \rightarrow \boxed{h[n]} \rightarrow$

$$y[0] = \sum_{k=-\infty}^{\infty} h[k]x[-k] = \sum_{k=-\infty}^{\infty} \frac{h[k]h[k]}{|h[k]|} = \sum_{k=-\infty}^{\infty} |h[k]| = \infty$$



Stability of LTI Systems

□ The continuous case

$$\begin{aligned} |y(t)| &= \left| \int_{-\infty}^{\infty} h(\tau) x(t - \tau) d\tau \right| \\ &\leq \int_{-\infty}^{\infty} |h(\tau)| |x(t - \tau)| d\tau \\ &\leq B_x \int_{-\infty}^{\infty} |h(\tau)| d\tau \\ &< \infty \end{aligned}$$



Example

- Consider an LTI discrete-time system with an impulse response $h[n] = \alpha^n \mu[n]$
- For this system:

$$S = \sum_{n=-\infty}^{\infty} |\alpha^n| \mu[n] = \sum_{n=0}^{\infty} |\alpha|^n = \frac{1}{1 - |\alpha|}, \text{ if } |\alpha| < 1$$

- Therefore $S < \infty$ if $|\alpha| < 1$, i.e, the system is **stable**
- If $|\alpha| = 1$, the system is **unstable**



The Unit Step Response

- The unit step response, $s(t)$ or $s[n]$, corresponding to the output when $x[n]=u[n]$ or $x(t)=u(t)$

$$s[n] = u[n] * h[n] = \sum_{k=-\infty}^{\infty} h[k]u[n-k] = \sum_{k=-\infty}^n h[k]$$

$$h[n] = s[n] - s[n-1]$$

$$s(t) = u(t) * h(t) = \int_{-\infty}^{\infty} h(\tau)u(t-\tau)d\tau = \int_{-\infty}^t h(\tau)d\tau$$

$$h(t) = \frac{ds(t)}{dt} = s'(t)$$



CT-LTI System: Differential Equation

□ A first-order example

$$\frac{dy(t)}{dt} + 2y(t) = x(t)$$



CT-LTI System: Differential Equation

□ General case

$$\sum_{k=0}^N a_k \frac{d^k y(t)}{dt^k} = \sum_{k=0}^m b_k \frac{d^k x(t)}{dt^k}$$

□ A solution generally consist of a particular solution and a homogeneous solution, i.e.

$$y(t) = y_p(t) + y_h(t)$$

□ Initial rest used as auxiliary condition, that is $x(t)=0$ for $t \leq t_0$, i.e.,

$$y(t_0) = \frac{dy(t_0)}{dt} = \dots = \frac{d^{N-1} y(t_0)}{dt^{N-1}} = 0$$



DT-LTI System: Difference Equation

□ General case

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

□ $y[n]$ can be computed recursively

$$a_0 y[n] = \sum_{k=0}^M b_k x[n-k] - \sum_{k=1}^N a_k y[n-k]$$

□ When a_0 is normalized to $a_0 = 1$

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



Example

□ Consider $y[n] - \frac{1}{2} y[n-1] = x[n]$, determine $h[n]=?$

□ A: the system is initially reset, i.e.,
 $y[n]=0$ for $n \leq -1$, and let $x[n] = \delta[n]$

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

Impulse response

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

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

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

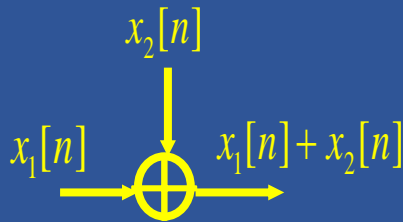
⋮

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



Block Diagrams

□ Basic operations



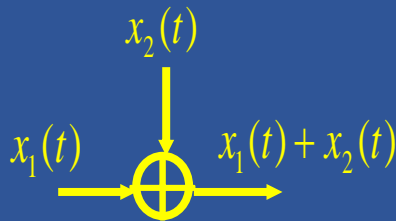
(a) An adder



(b) multiplication by a coefficient



(c) a unit delay



(d) An adder



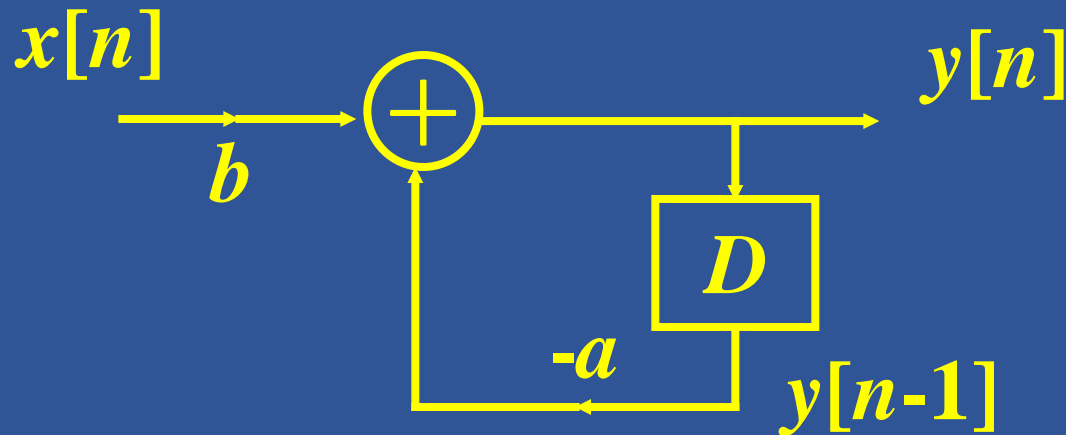
(e) multiplication by a coefficient



(f) a differentiator

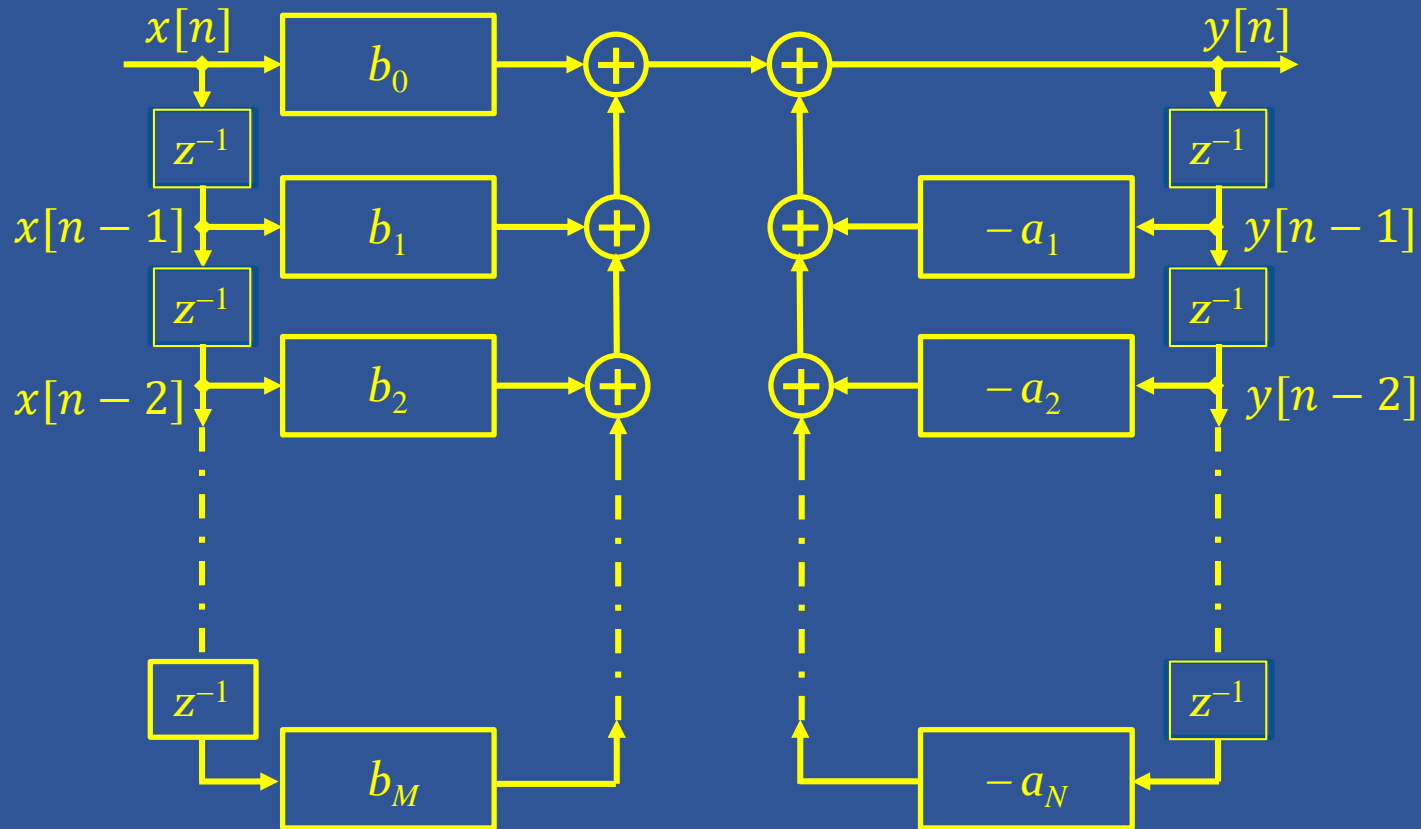
Block Diagrams: Example

□ Block diagram for $y[n] + ay[n-1] = bx[n]$



Block Diagrams: Example

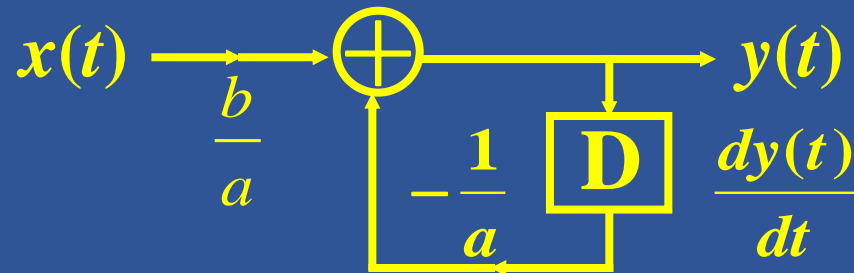
$$y[n] = \sum_{k=0}^M b_k x[n-k] - \sum_{k=1}^N a_k y[n-k]$$



Block Diagrams: Example

□ Block diagram for $\frac{dy(t)}{dt} + ay(t) = bx(t)$

□ A: rewrite the equation to $y(t) = -\frac{1}{a} \frac{dy(t)}{dt} + \frac{b}{a} x(t)$



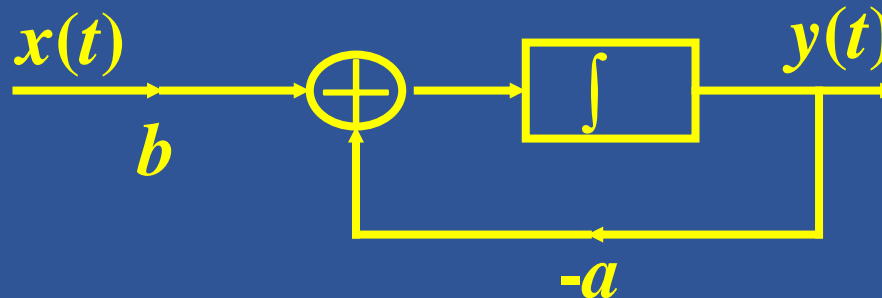
Another Implementation

□ The system can be implemented using an integrator

□ A:
$$\frac{dy(t)}{dt} + ay(t) = bx(t) \rightarrow \frac{dy(t)}{dt} = bx(t) - ay(t)$$

integrating from $-\infty$ to t :

$$y(t) = \int_{-\infty}^t [bx(\tau) - ay(\tau)] d\tau$$



Classification of DT LTI System

- If $h[n]$ is of **finite** length, then it is known as a **finite impulse response (FIR)** system

$$y[n] = \sum_{k=N_1}^{N_2} h[k]x[n-k]$$

➤ Examples: moving averaged filter

$$y[n] = \sum_{k=0}^4 \frac{1}{5} x[n-k]$$



Classification of DT LTI System

- If $h[n]$ is of **infinite** length, then it is known as a **infinite impulse response (IIR)** system

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

- If causal (system causal, input sequence causal),

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

- Example: accumulator

$$y[n] = \sum_{l=-\infty}^n x[l] = y[n-1] + x[n]$$

