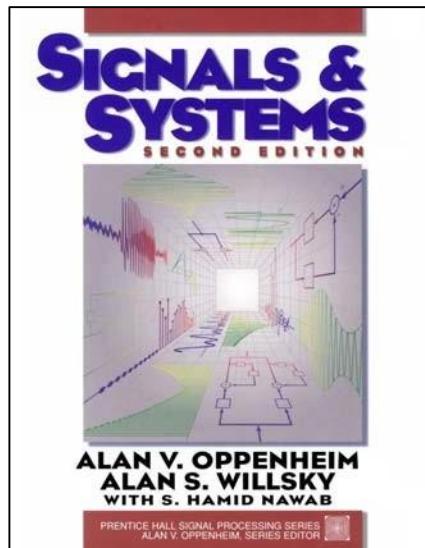


Spring 2010

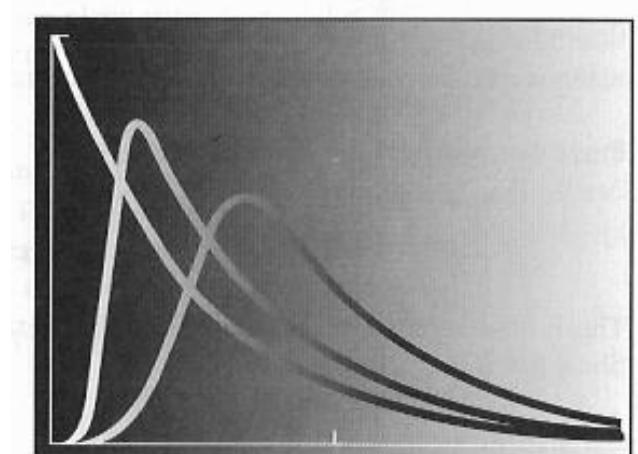
信號與系統 Signals and Systems

Chapter SS-2 Linear Time-Invariant Systems



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NTU-EE
Feb10 – Jun10

Figures and images used in these lecture notes are adopted from
"Signals & Systems" by Alan V. Oppenheim and Alan S. Willsky, 1997



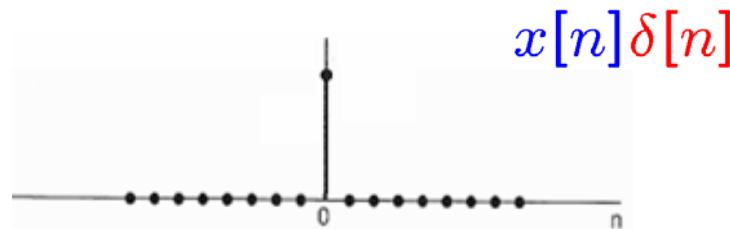
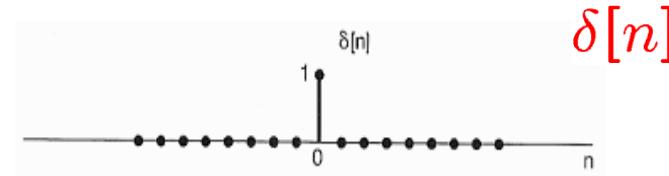
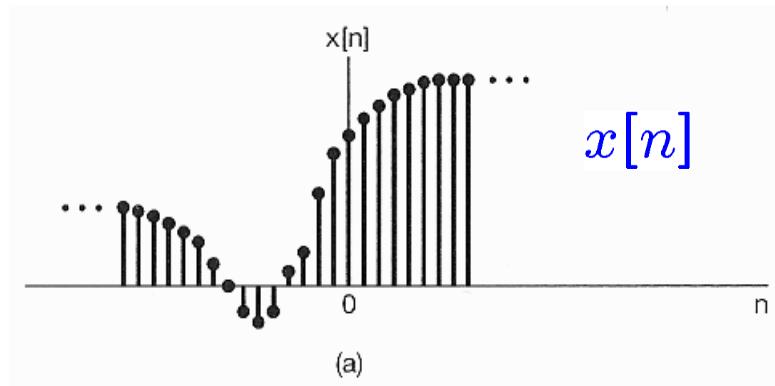
- Discrete-Time Linear Time-Invariant Systems
 - The convolution sum
- Continuous-Time Linear Time-Invariant Systems
 - The convolution integral
- Properties of Linear Time-Invariant Systems
- Causal Linear Time-Invariant Systems
Described by Differential & Difference Equations
- Singularity Functions

$$x[n] \rightarrow \boxed{h[n]} \rightarrow y[n] \qquad x(t) \rightarrow \boxed{h(t)} \rightarrow y(t)$$

- Sample by Unit Impulse

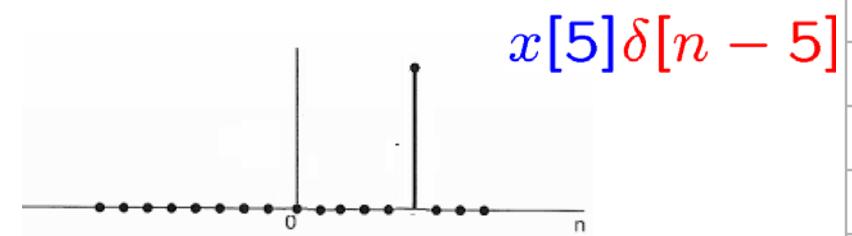
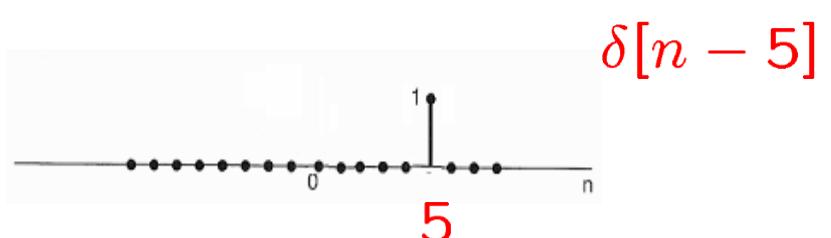
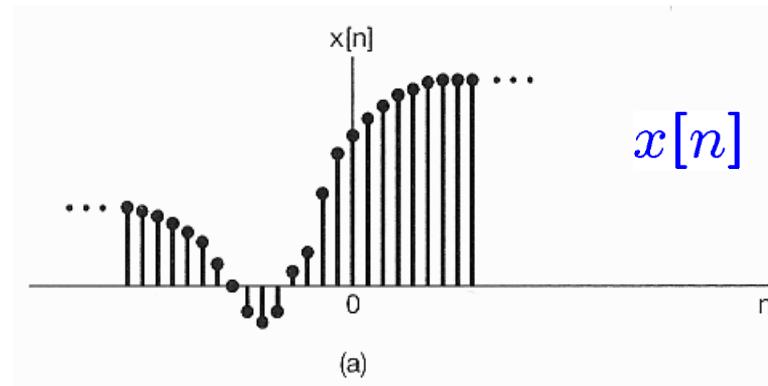
- For $x[n]$

$$x[n]\delta[n] = x[0]\delta[n]$$

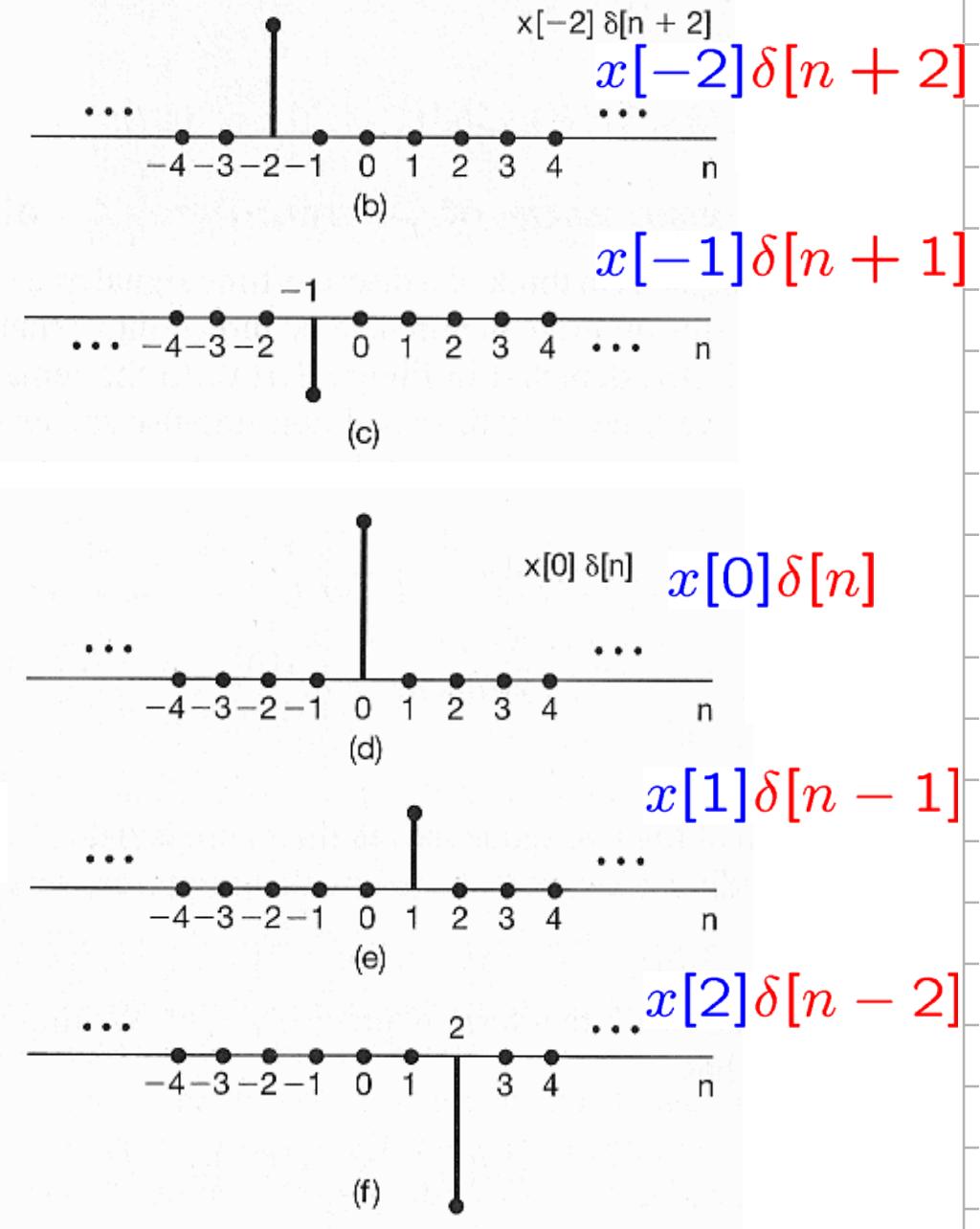
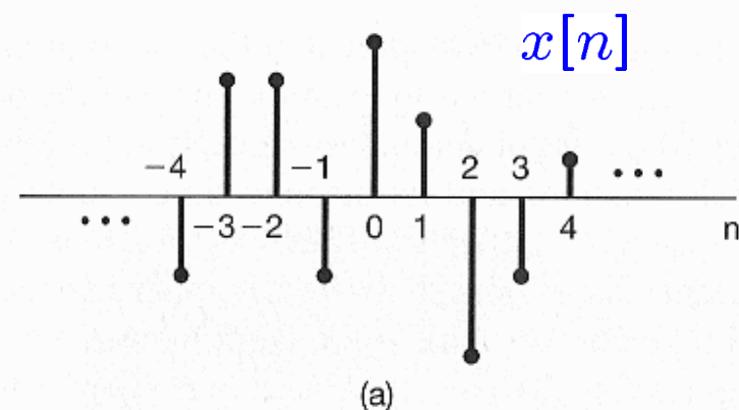


- More generally,

$$x[n]\delta[n - n_0] = x[n_0]\delta[n - n_0]$$



- Representation of DT Signals by Impulses



▪ Representation of DT Signals by Impulses:

- More generally,

$$x[n] = \dots + x[-3]\delta[n+3] + x[-2]\delta[n+2]$$

$$+ x[-1]\delta[n+1] + x[0]\delta[n] + x[1]\delta[n-1]$$

$$+ x[2]\delta[n-2] + x[3]\delta[n-3] + \dots$$

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

- The **sifting property** of the DT unit impulse
- $x[n]$ = a **superposition** of scaled versions of **shifted unit impulses** $\delta[n-k]$

■ DT Unit Impulse Response & Convolution Sum:

input → Linear System → output

$\delta[n]$ → Linear System → $h_0[n]$

$\delta[n - 1]$ → Linear System → $h_1[n]$

$\delta[n - 2]$ → Linear System → $h_2[n]$

⋮

$\delta[n - k]$ → Linear System → $h_k[n]$

■ DT Unit Impulse Response & Convolution Sum:

$$x[n] \rightarrow \text{Linear System} \rightarrow y[n]$$

$$x[0] \cdot \delta[n] \rightarrow \text{Linear System} \rightarrow h_0[n] \cdot x[0]$$

$$x[1] \cdot \delta[n - 1] \rightarrow \text{Linear System} \rightarrow h_1[n] \cdot x[1]$$

$$x[2] \cdot \delta[n - 2] \rightarrow \text{Linear System} \rightarrow h_2[n] \cdot x[2]$$

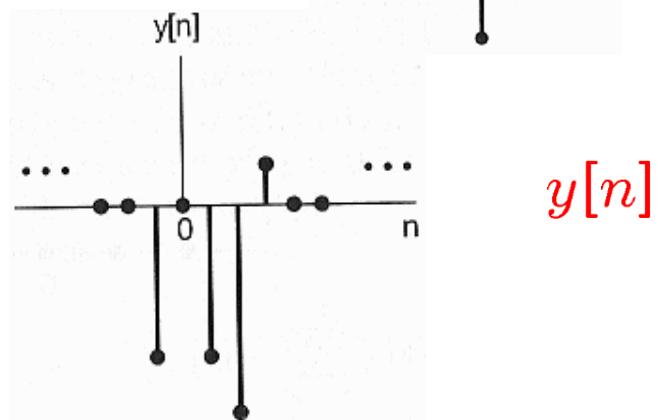
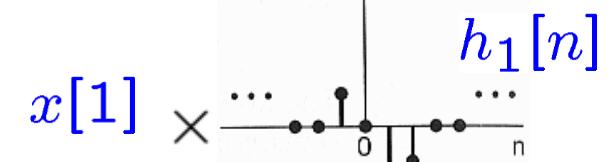
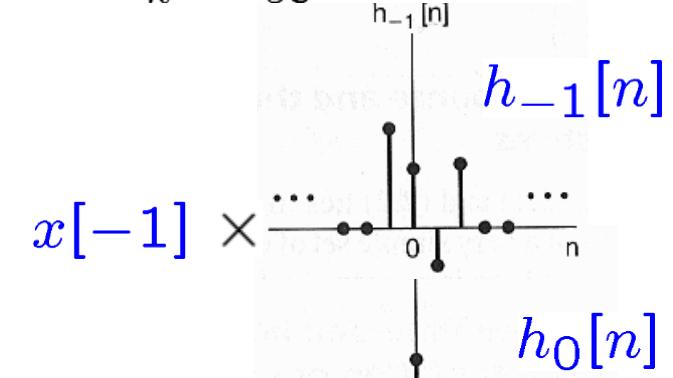
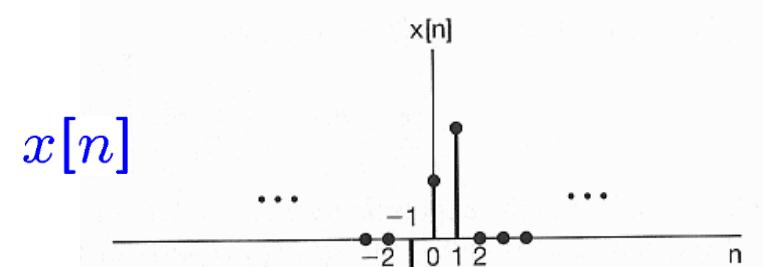
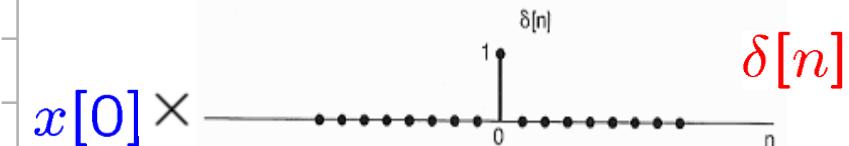
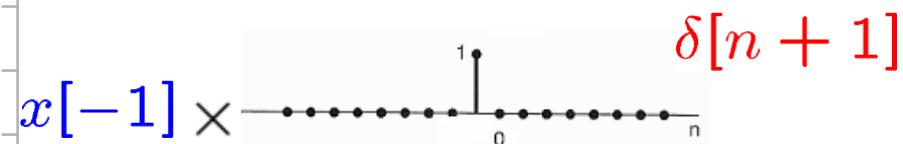
 \vdots

$$x[k] \cdot \delta[n - k] \rightarrow \text{Linear System} \rightarrow h_k[n] \cdot x[k]$$

$$x[n] = \sum_{k=-\infty}^{+\infty} x[k] \delta[n - k] \quad \Rightarrow \quad y[n] = \sum_{k=-\infty}^{+\infty} x[k] h_k[n]$$

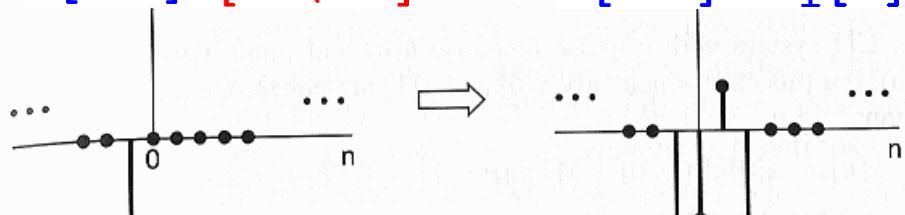
DT LTI Systems: Convolution Sum

$$x[n] = \sum_{k=-\infty}^{+\infty} x[k] \delta[n-k] \implies y[n] = \sum_{k=-\infty}^{+\infty} x[k] h_k[n]$$



$$x[n] = \sum_{k=-\infty}^{+\infty} x[k]\delta[n-k] \implies y[n] = \sum_{k=-\infty}^{+\infty} x[k]h_k[n]$$

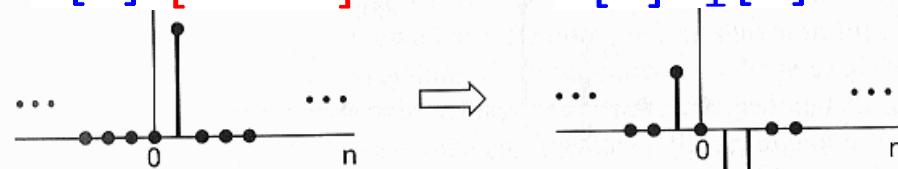
$$x[-1]\delta[n+1] \quad x[-1]h_{-1}[n]$$



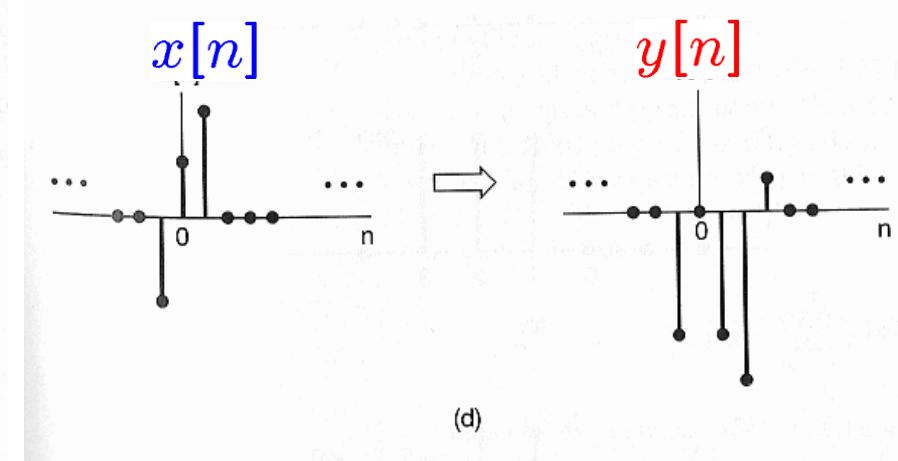
$$x[0]\delta[n]$$



$$x[1]\delta[n-1]$$



(c)



(d)

$x[n] \rightarrow$ Linear System $\rightarrow y[n]$

- If the linear system (L) is also time-invariant (TI)

- Then,

$$h_k[n] = h_0[n - k] = h[n - k]$$

- Hence, for an LTI system,

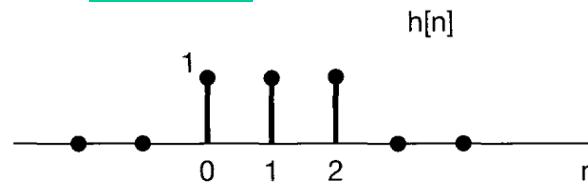
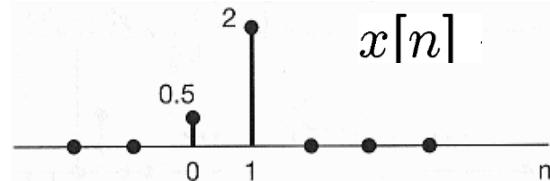
$$y[n] = \sum_{k=-\infty}^{+\infty} x[k]h[n - k] = \sum_{k=-\infty}^{+\infty} x[n - k]h[k]$$

- Known as the convolution of $x[n]$ & $h[n]$
 - Referred as the convolution sum or superposition sum

- Symbolically, $y[n] = x[n] * h[n] = h[n] * x[n]$

■ Example 2.1:

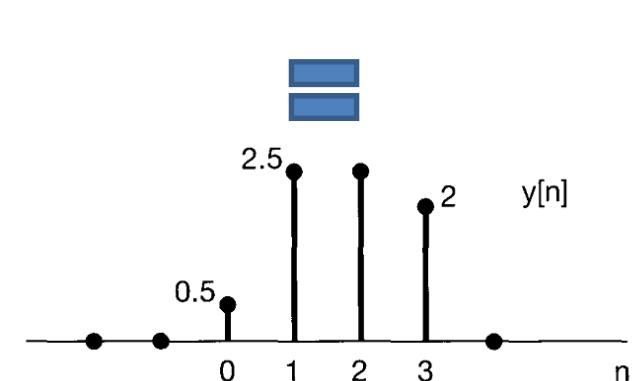
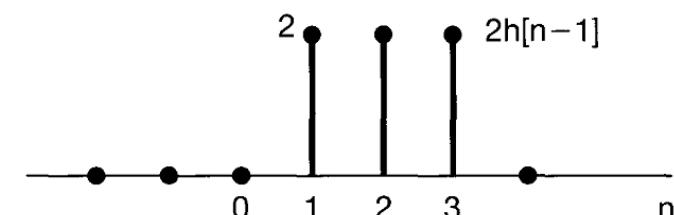
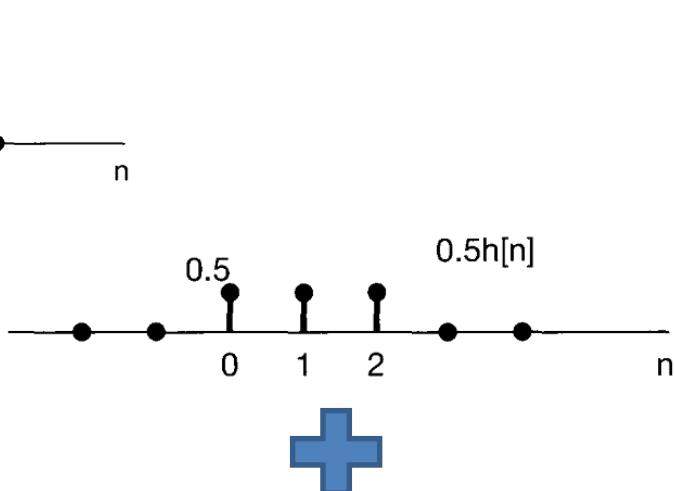
$$x[n] \longrightarrow h[n] \longrightarrow y[n]$$



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

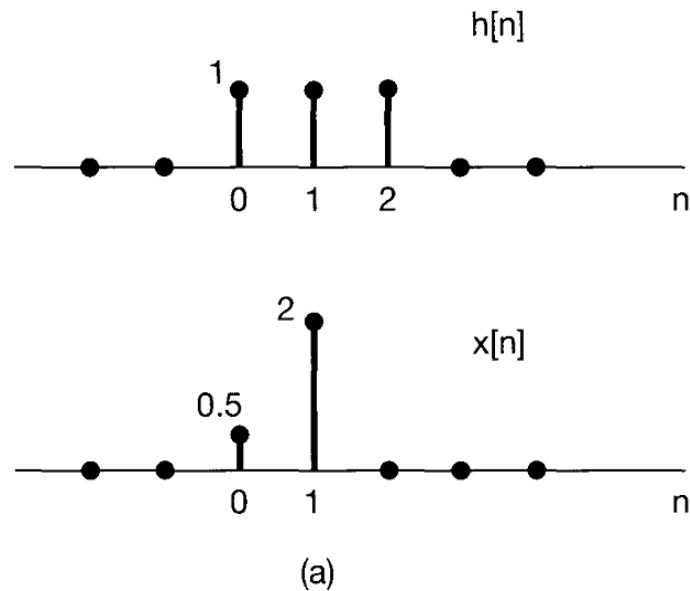
$$y[n] = x[0]h[n-0] + x[1]h[n-1]$$

$$= 0.5h[n] + 2h[n-1]$$

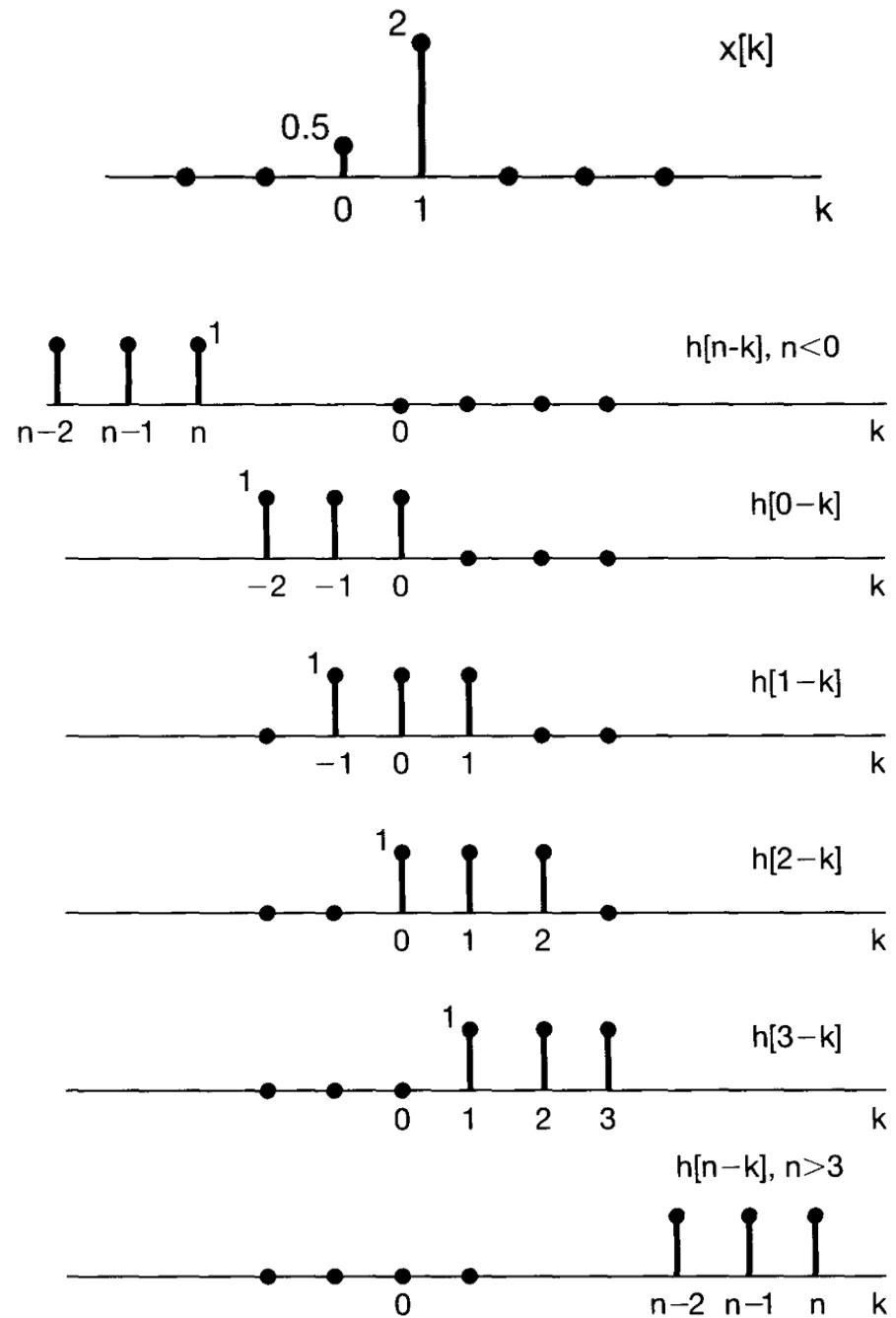


■ Example 2.2:

The problem is as Example 2.1, but , we consider k as the independent variable!

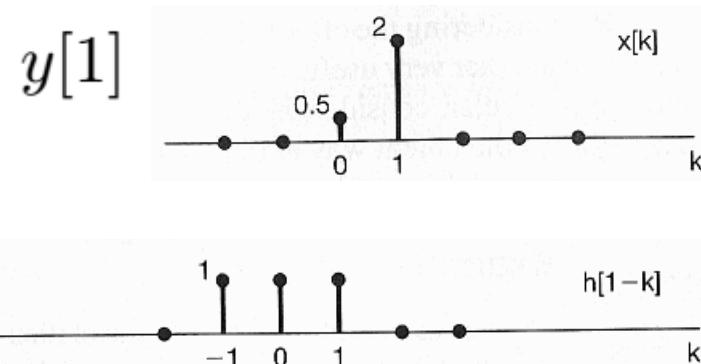
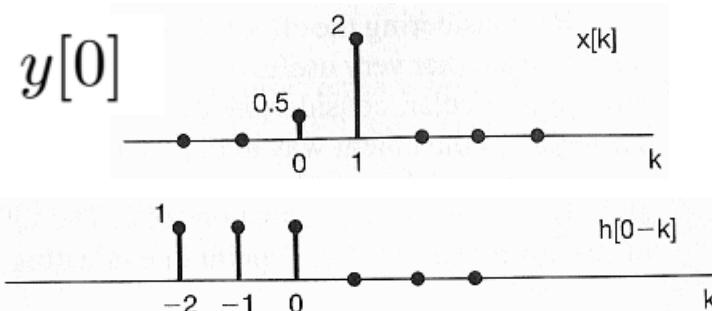
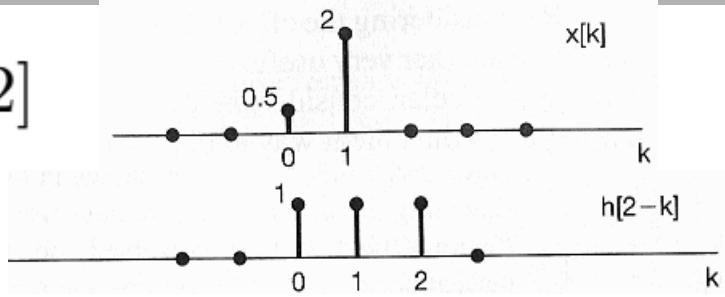
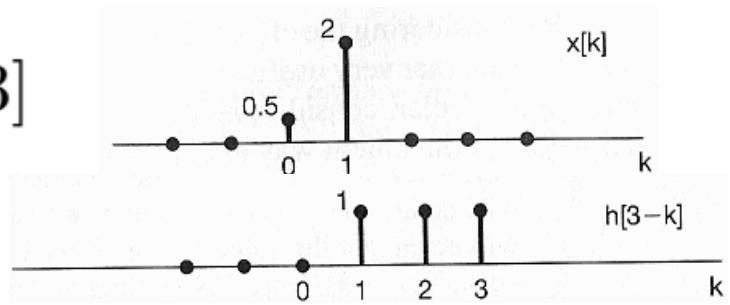
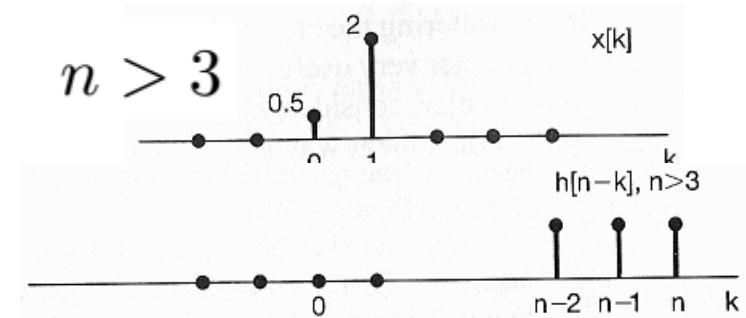
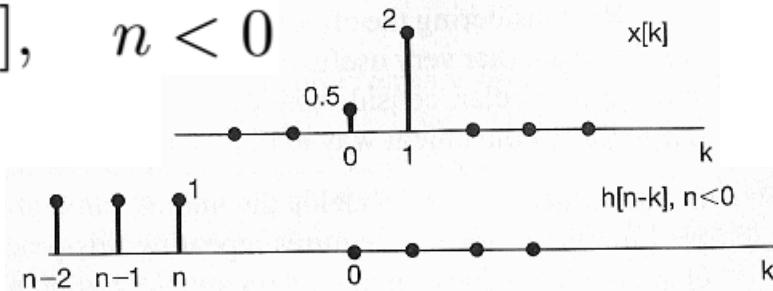


$$y[n] = \sum_{k=-\infty}^{+\infty} x[k] h[n-k]$$



■ Example 2.2:

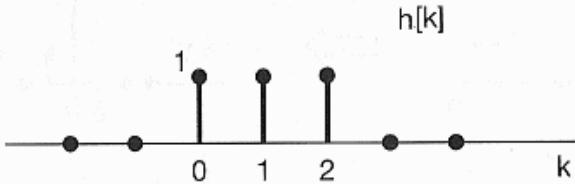
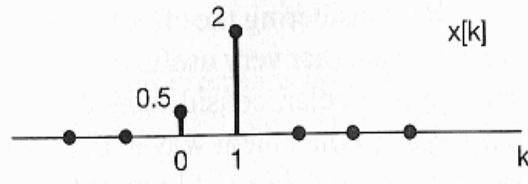
$$y[n] = \sum_{k=-\infty}^{+\infty} x[k]h[n-k]$$

 $y[2]$  $y[3]$  $y[n], n > 3$  $y[n], n < 0$ 

Example 2.2:

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

$$x[n] \longrightarrow h[n] \longrightarrow y[n]$$



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

$$= \cdots + x[-1] h[1] + x[0] h[0] + x[1] h[-1] + x[2] h[-2] + \cdots = 0.5$$

$$y[1] = \sum_{k=-\infty}^{+\infty} x[k] h[1 - k] = 2.5$$

$$= \cdots + x[-1] h[2] + x[0] h[1] + x[1] h[0] + x[2] h[-1] + \cdots = 2.5$$

$$y[2] = \sum_{k=-\infty}^{+\infty} x[k] h[2 - k] = 2.5$$

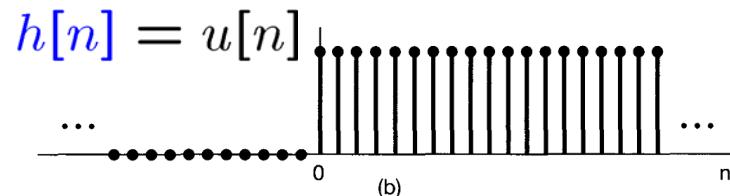
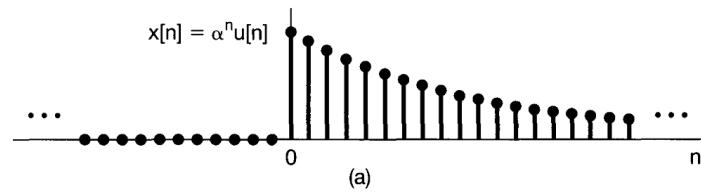
$$y[n] = 0 \text{ for } n < 0$$

$$y[3] = \sum_{k=-\infty}^{+\infty} x[k] h[3 - k] = 2.0$$

$$y[n] = 0 \text{ for } n > 3$$

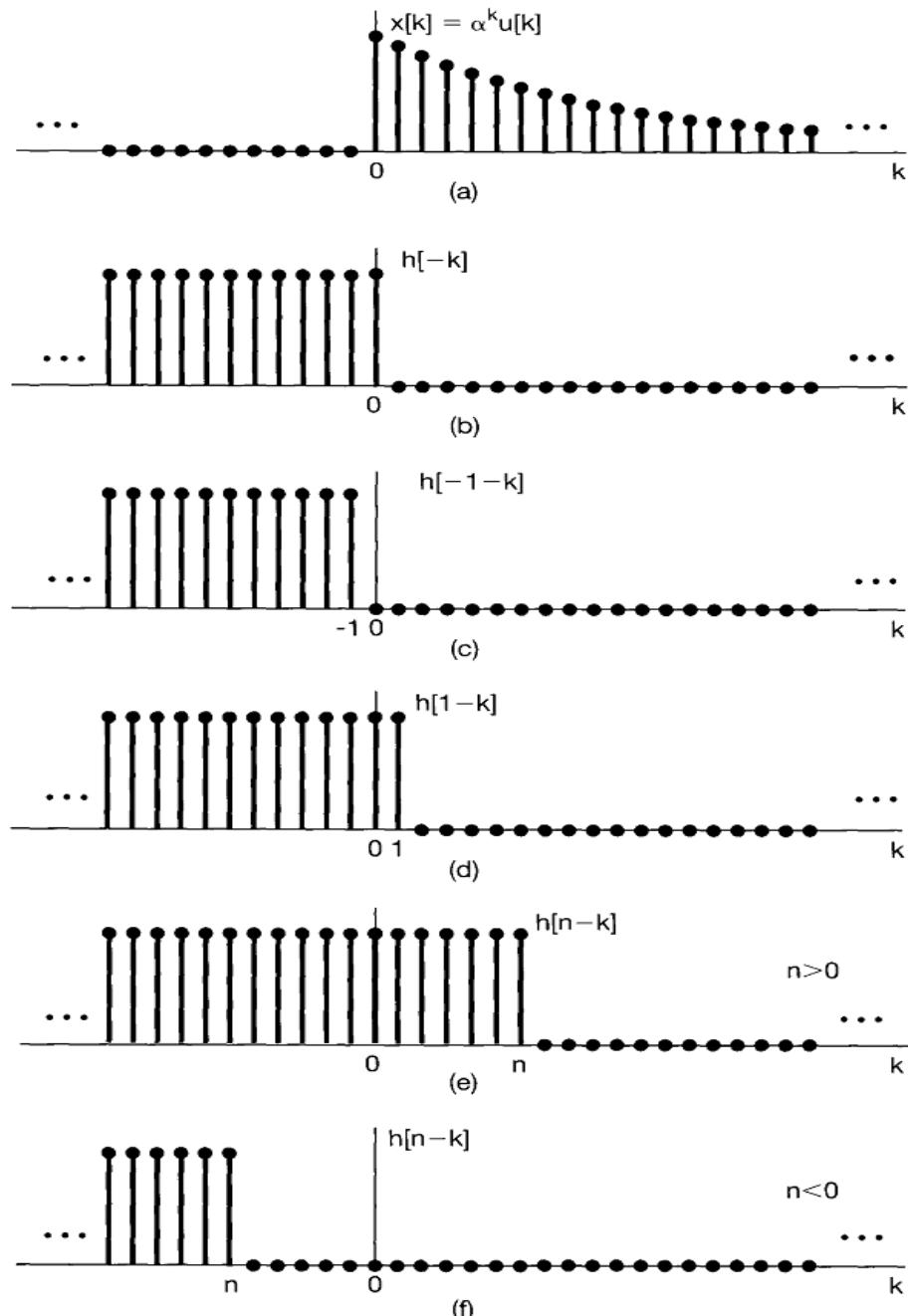
■ Example 2.3:

$$x[n] = \alpha^n u[n], \quad 0 < \alpha < 1$$



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

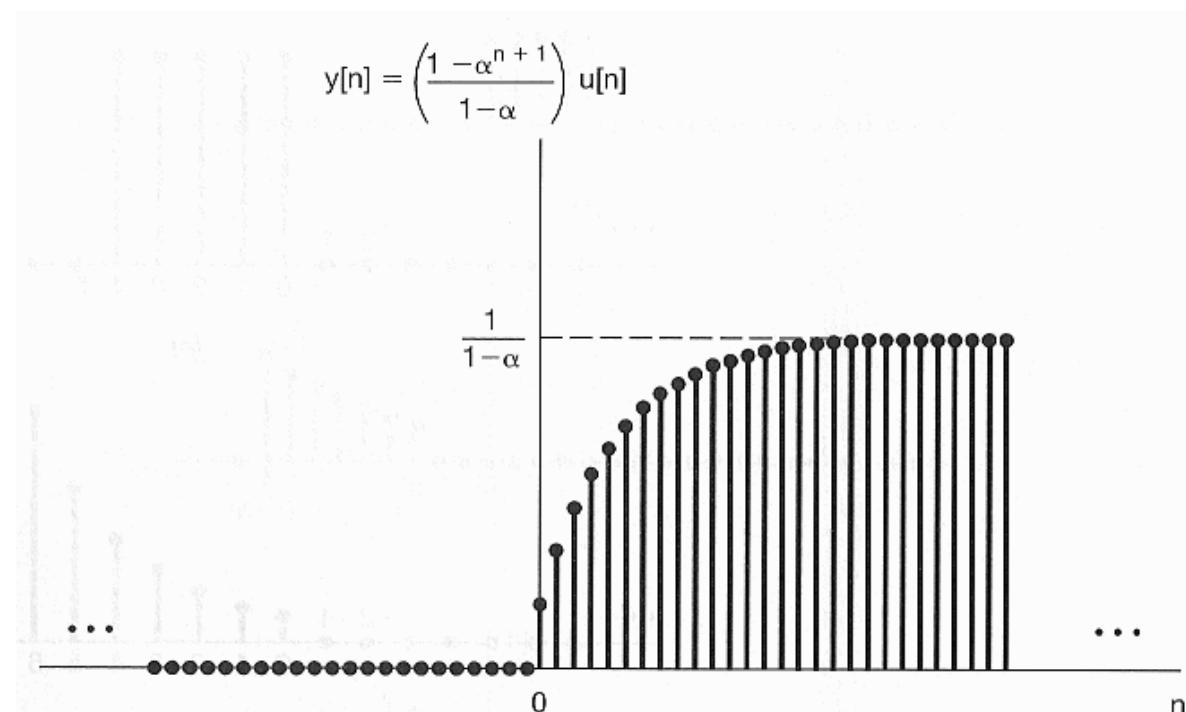
We consider k as
the independent variable!



■ Example 2.3:

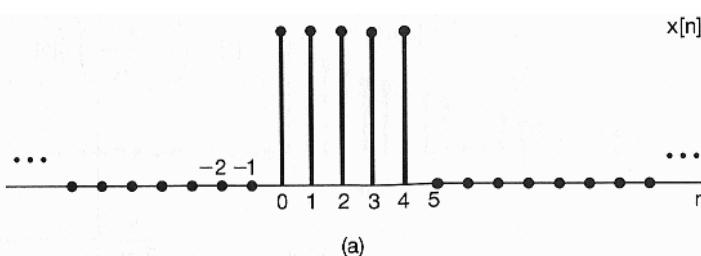
$$y[n] = \sum_{k=-\infty}^{+\infty} x[k] h[n-k]$$

$$x[k] h[n-k] = \begin{cases} \alpha^k, & 0 \leq k \leq n \\ 0, & \text{otherwise} \end{cases} \quad \text{for } n \geq 0, \Rightarrow y[n] = \sum_{k=0}^n \alpha^k = \frac{1 - \alpha^{n+1}}{1 - \alpha}$$

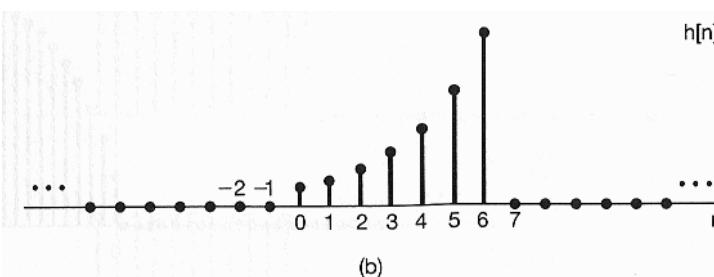


■ Example 2.4:

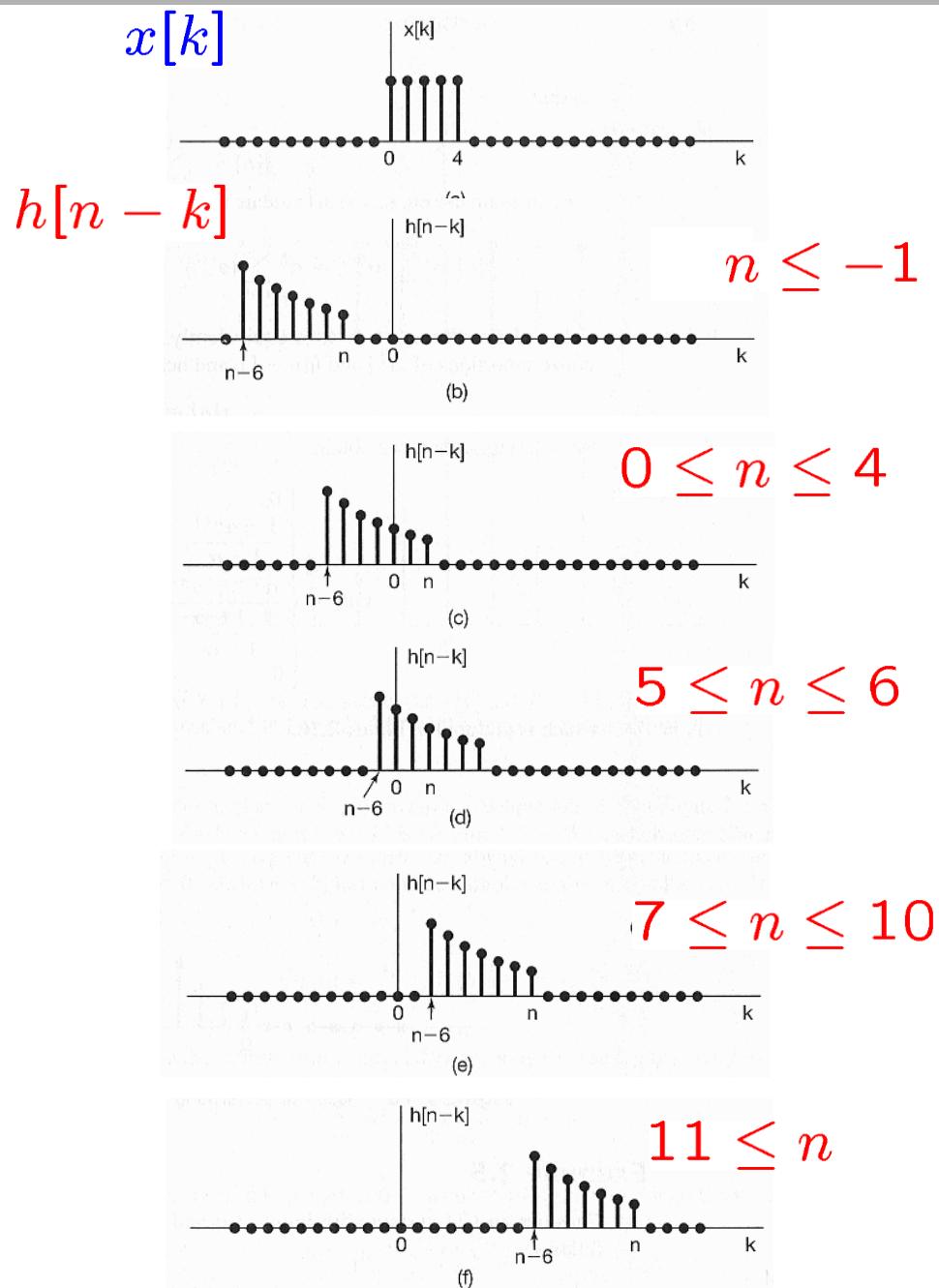
$$x[n] = 1, \quad 0 \leq n \leq 4$$



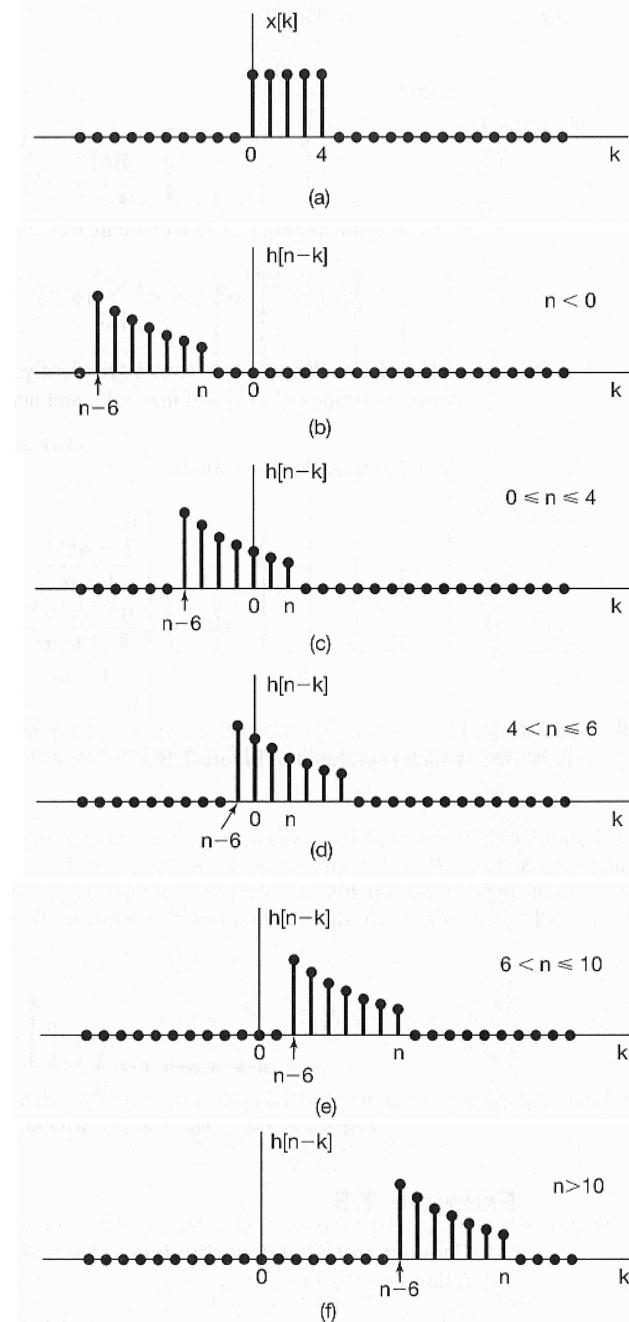
$$h[n] = \alpha^n, \quad 0 \leq n \leq 6$$



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



DT LTI Systems: Convolution Sum



for $n < 0$, $x[k] h[n - k] = 0 \Rightarrow y[n] = 0$

for $0 \leq n \leq 4$, $x[k] h[n - k] = \begin{cases} \alpha^{n-k}, & 0 \leq k \leq n \\ 0, & \text{otherwise} \end{cases}$

$$\Rightarrow y[n] = \sum_{k=0}^n \alpha^{n-k} = \frac{1 - \alpha^{n+1}}{1 - \alpha}$$

for $4 < n \leq 6$, $x[k] h[n - k] = \begin{cases} \alpha^{n-k}, & 0 \leq k \leq 4 \\ 0, & \text{otherwise} \end{cases}$

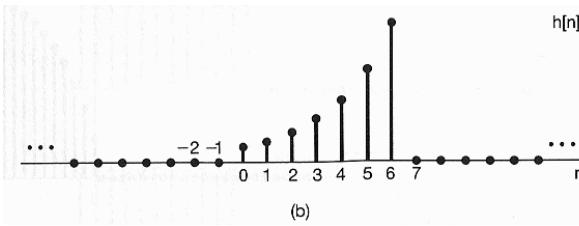
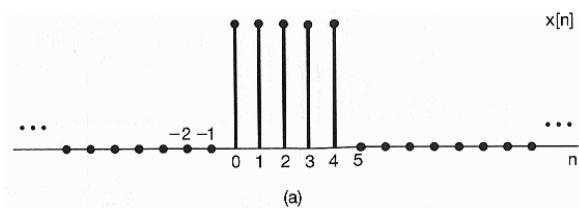
$$\Rightarrow y[n] = \sum_{k=0}^4 \alpha^{n-k} = \frac{\alpha^{n-4} - \alpha^{n+1}}{1 - \alpha}$$

for $6 < n \leq 10$, $x[k] h[n - k] = \begin{cases} \alpha^{n-k}, & (n - 6) \leq k \leq 4 \\ 0, & \text{otherwise} \end{cases}$

$$\Rightarrow y[n] = \sum_{k=n-6}^4 \alpha^{n-k} = \frac{\alpha^{n-4} - \alpha^7}{1 - \alpha}$$

for $n > 10$, $y[n] = 0$

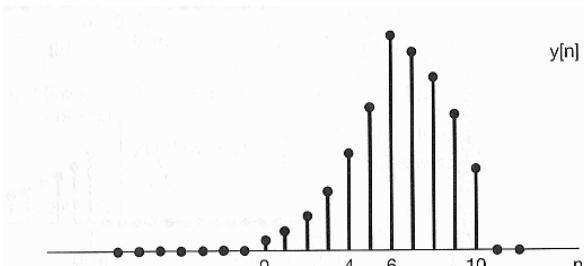




$$x[n] \longrightarrow h[n] \longrightarrow y[n]$$

$$x[n] = 1, \quad 0 \leq n \leq 4$$

$$h[n] = \alpha^n, \quad 0 \leq n \leq 6$$



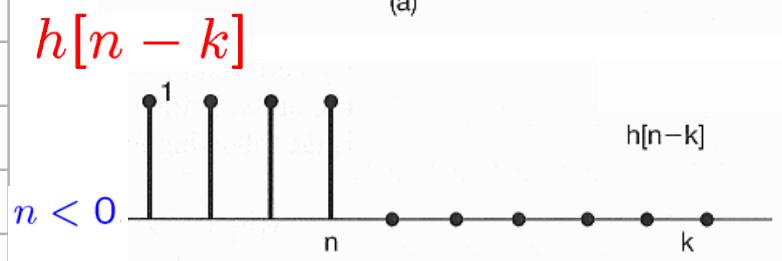
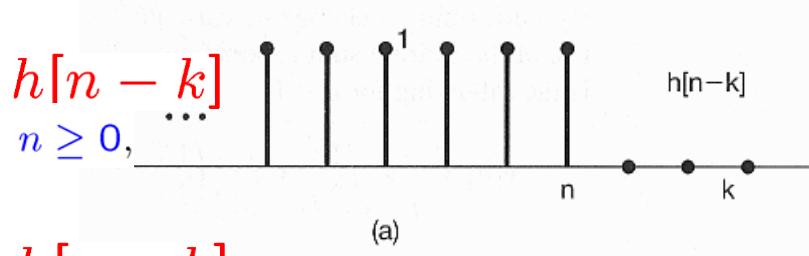
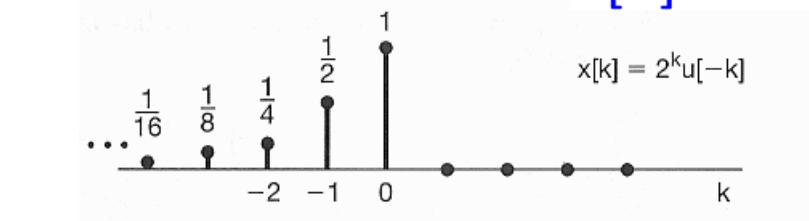
$$y[n] = \begin{cases} 0, & n < 0 \\ \frac{1-\alpha^{n+1}}{1-\alpha}, & 0 \leq n \leq 4 \\ \frac{\alpha^{n-4}-\alpha^{n+1}}{1-\alpha}, & 4 < n \leq 6 \\ \frac{\alpha^{n-4}-\alpha^7}{1-\alpha}, & 6 < n \leq 10 \\ 0, & 10 < n \end{cases}$$

■ Example 2.5:

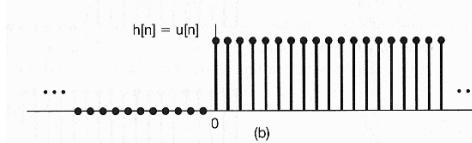
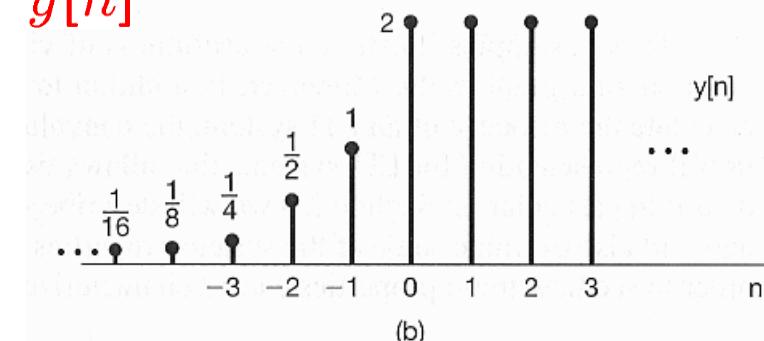
$$x[n] = 2^n u[-n]$$

$$x[n] \longrightarrow h[n] \longrightarrow y[n]$$

$$h[n] = u[n]$$



$y[n]$



$$\text{for } n \geq 0, \quad y[n] = \sum_{k=-\infty}^0 x[k] h[n-k] = \sum_{k=-\infty}^0 2^k$$

$$= \sum_{r=0}^{\infty} \left(\frac{1}{2}\right)^r = \frac{1}{1 - (1/2)} = 2$$

$$\text{for } n < 0, \quad y[n] = \sum_{k=-\infty}^n x[k] h[n-k] = \sum_{k=-\infty}^n 2^k$$

$$= \sum_{l=-n}^{\infty} \left(\frac{1}{2}\right)^l = \sum_{m=0}^{\infty} \left(\frac{1}{2}\right)^{m-n}$$

$$= \left(\frac{1}{2}\right)^{-n} \sum_{m=0}^{\infty} \left(\frac{1}{2}\right)^m = 2^n \cdot 2 = 2^{n+1}$$

- Discrete-Time Linear Time-Invariant Systems

- The convolution sum

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

- Continuous-Time Linear Time-Invariant Systems

- The convolution integral

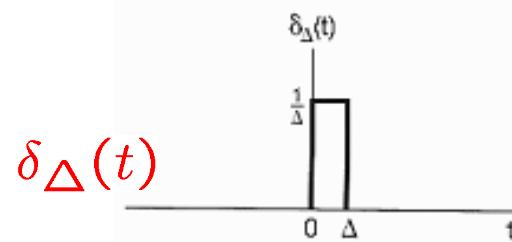
- Properties of Linear Time-Invariant Systems

- Causal Linear Time-Invariant Systems

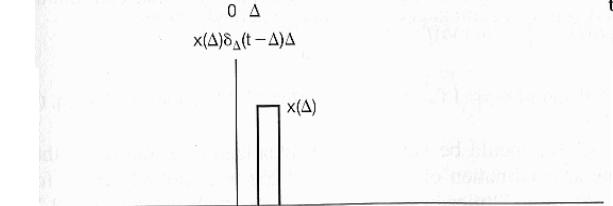
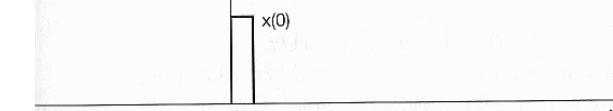
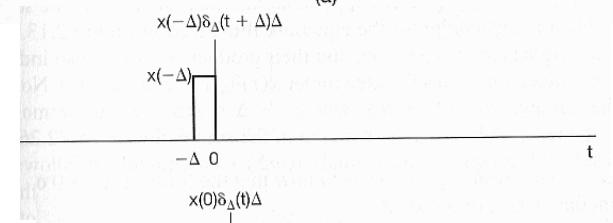
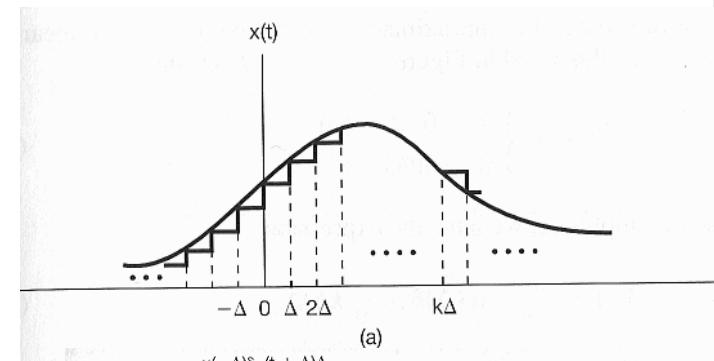
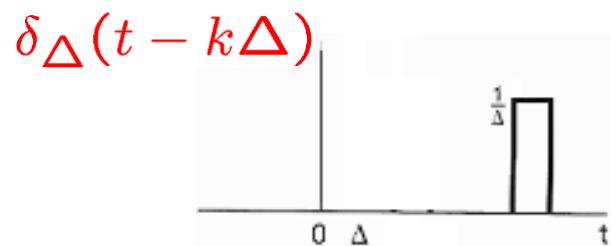
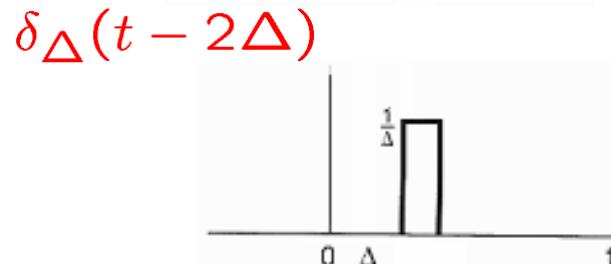
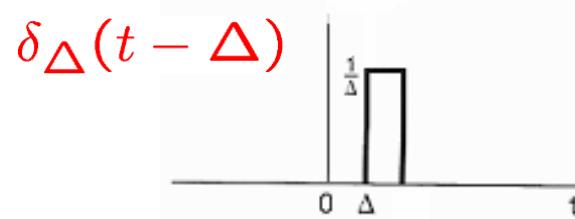
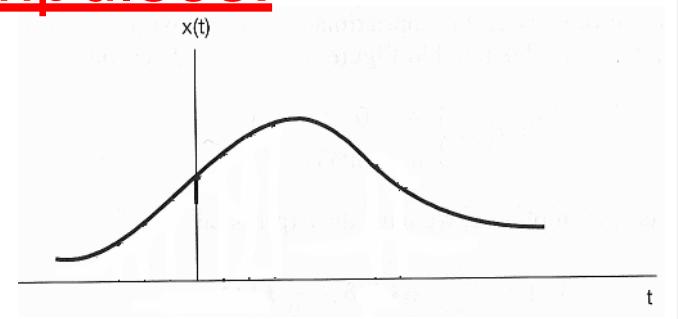
Described by Differential & Difference Equations

- Singularity Functions

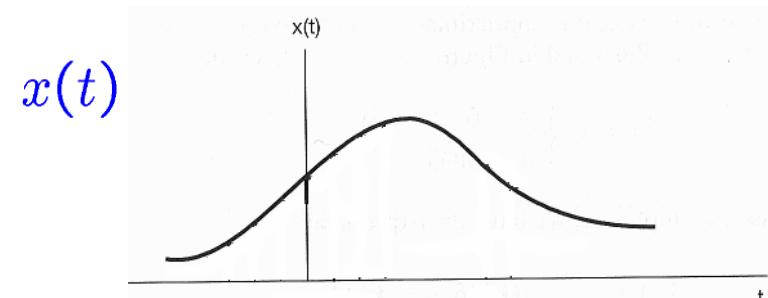
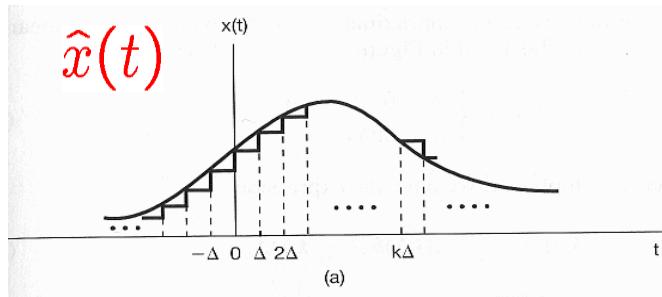
■ Representation of CT Signals by Impulses:



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



■ Representation of CT Signals by Impulses:



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

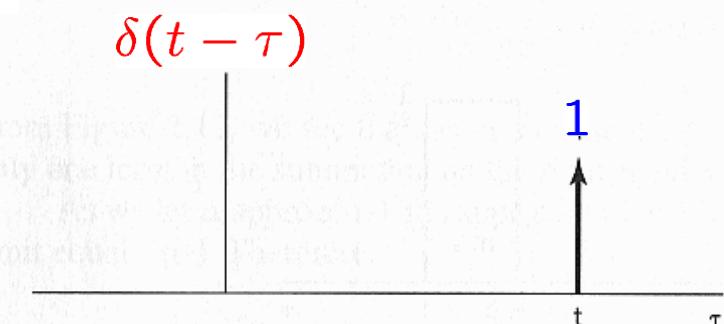
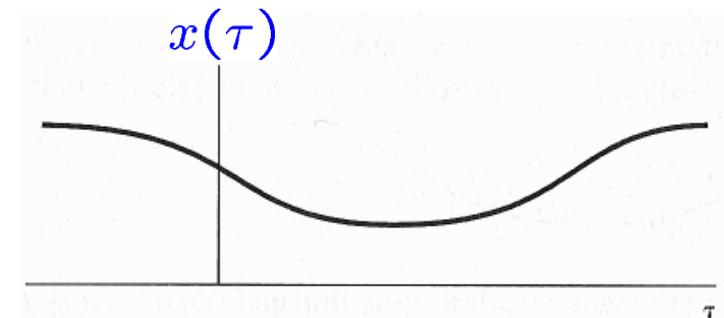
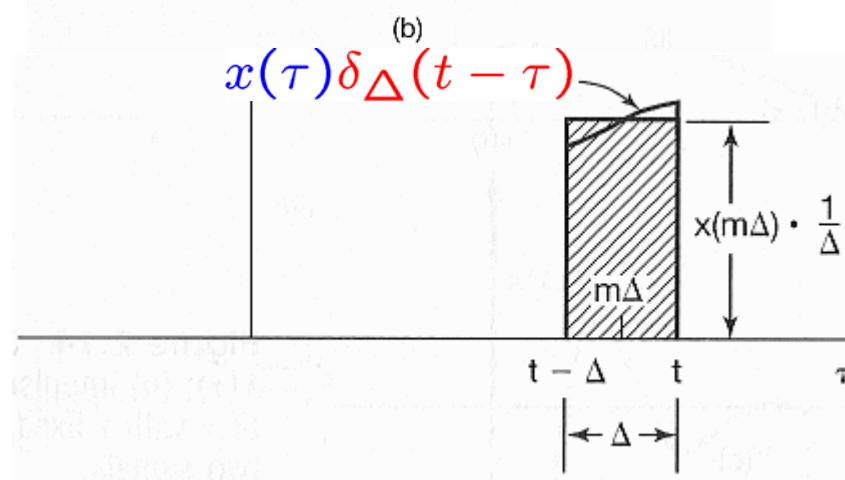
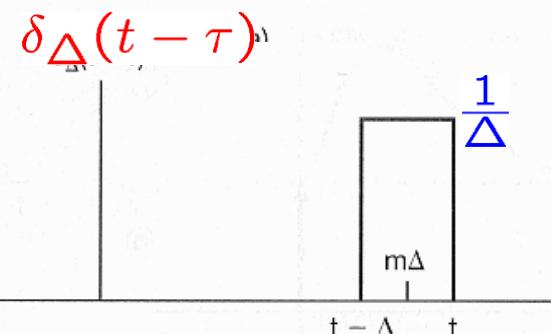
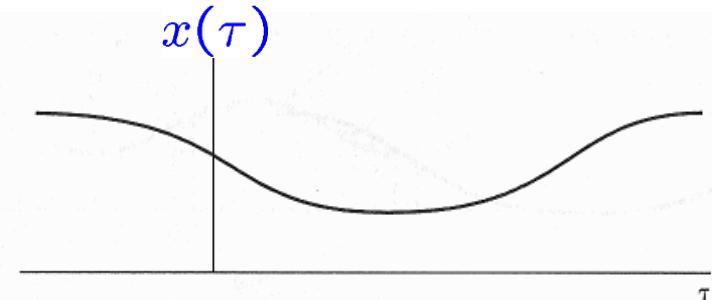
$$x(t) = \lim_{\Delta \rightarrow 0} \sum_{k=-\infty}^{+\infty} x(k\Delta) \delta_\Delta(t - k\Delta) \Delta$$

$$\int_{-\infty}^{+\infty} f(\tau) d\tau = \lim_{\Delta \rightarrow 0} \sum_{k=-\infty}^{k=+\infty} f(k\Delta) \Delta \quad \rightarrow \quad x(t) = \int_{-\infty}^{+\infty} x(\tau) \delta(t - \tau) d\tau$$

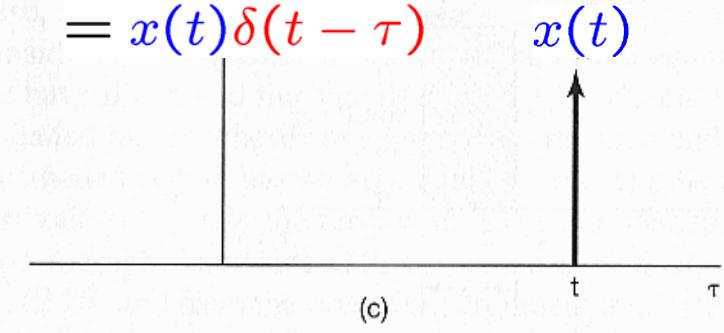
the **sifting property** of CT impulse

$x(t)$ = an integral of weighted, shifted impulses

■ Graphical interpretation:



$$\begin{aligned} & x(\tau)\delta(t - \tau) \\ &= x(t)\delta(t - \tau) \end{aligned}$$



(c)

■ CT Impulse Response & Convolution Integral:

input → Linear System → output

$\delta_{\Delta}(t) \rightarrow$ Linear System → $\hat{h}_{0\Delta}(t)$

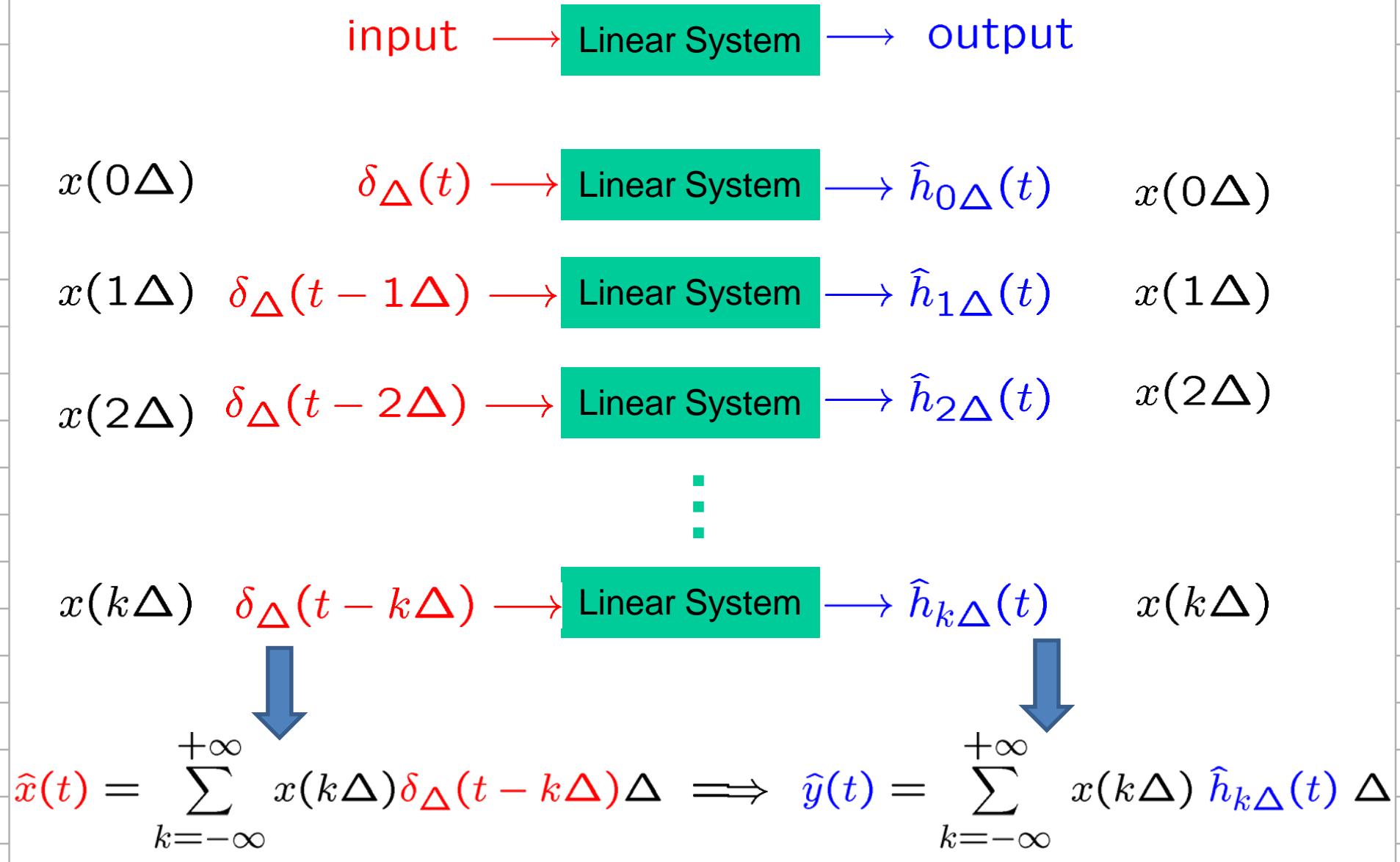
$\delta_{\Delta}(t - 1\Delta) \rightarrow$ Linear System → $\hat{h}_{1\Delta}(t)$

$\delta_{\Delta}(t - 2\Delta) \rightarrow$ Linear System → $\hat{h}_{2\Delta}(t)$

⋮

$\delta_{\Delta}(t - k\Delta) \rightarrow$ Linear System → $\hat{h}_{k\Delta}(t)$

■ CT Impulse Response & Convolution Integral:



■ CT Unit Impulse Response & Convolution Integral:

$$\hat{x}(t) = \sum_{k=-\infty}^{+\infty} x(k\Delta) \delta_\Delta(t - k\Delta) \Delta \longrightarrow \text{Linear System} \longrightarrow \hat{y}(t) = \sum_{k=-\infty}^{+\infty} x(k\Delta) \hat{h}_{k\Delta}(t) \Delta$$

$$x(t) = \lim_{\Delta \rightarrow 0} \sum_{k=-\infty}^{+\infty} x(k\Delta) \delta_\Delta(t - k\Delta) \longrightarrow \text{Linear System} \longrightarrow y(t) = \lim_{\Delta \rightarrow 0} \sum_{k=-\infty}^{+\infty} x(k\Delta) \hat{h}_{k\Delta}(t) \Delta$$

$$\delta(t - \tau) \longrightarrow \text{Linear System} \longrightarrow h_\tau(t)$$

$$x(t) \longrightarrow \text{Linear System} \longrightarrow y(t)$$

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

- If the linear system (L) is also time-invariant (TI)

- Then,

$$x(t) \longrightarrow h(t) \longrightarrow y(t)$$

$$h_\tau(t) = h_0(t - \tau) = h(t - \tau)$$

- Hence, for an LTI system,

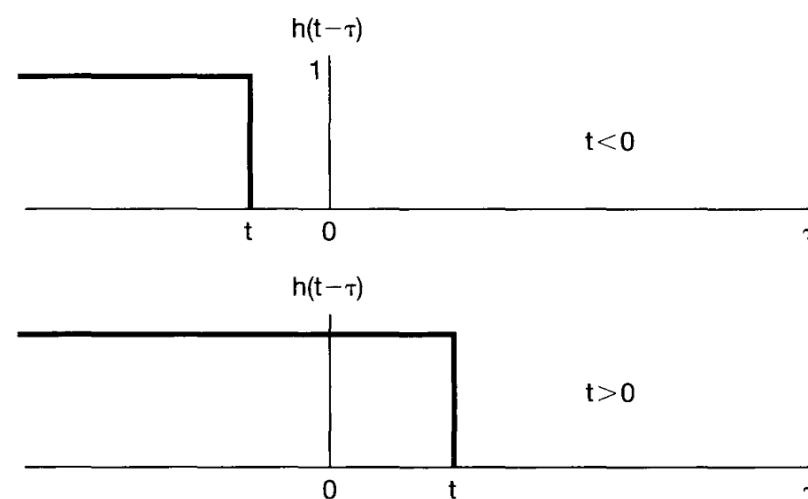
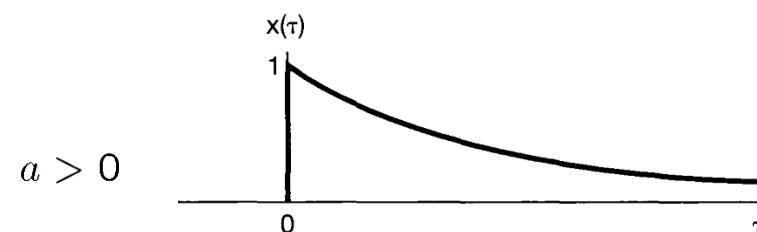
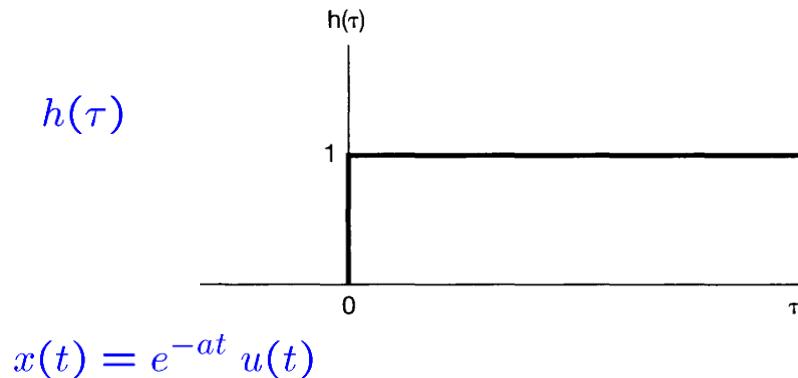
$$y(t) = \int_{-\infty}^{+\infty} x(\tau) h(t - \tau) d\tau = \int_{-\infty}^{+\infty} h(\tau) x(t - \tau) d\tau$$

- Known as the convolution of $x(t)$ & $h(t)$
 - Referred as the convolution integral or the superposition integral

- Symbolically,

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

■ Example 2.6: $y(t) = \int_{-\infty}^{+\infty} x(\tau) h(t - \tau) d\tau$



for $t < 0$, $x(\tau) h(t - \tau) = 0$

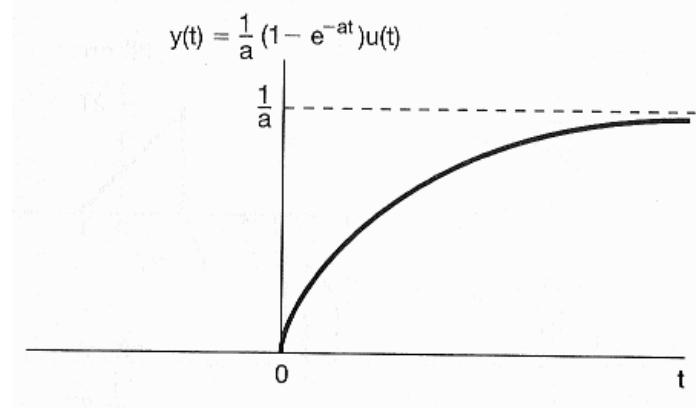
$$\Rightarrow y(t) = \int_{-\infty}^t 0 d\tau = 0$$

for $t \geq 0$, $x(\tau) h(t - \tau) = \begin{cases} e^{-a\tau}, & 0 \leq \tau \leq t \\ 0, & \text{otherwise} \end{cases}$

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

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

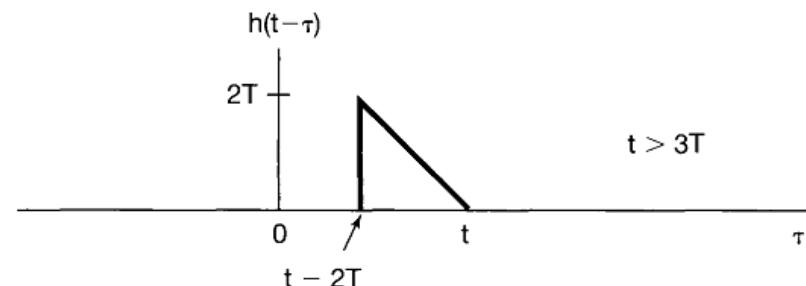
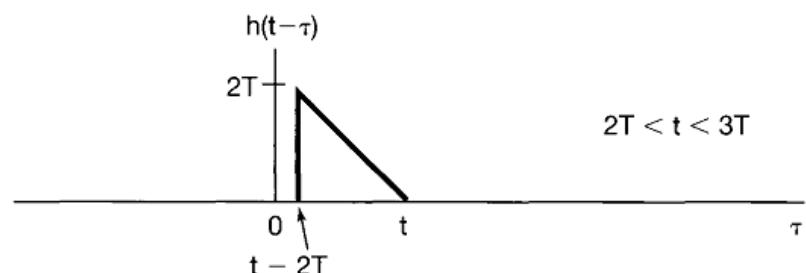
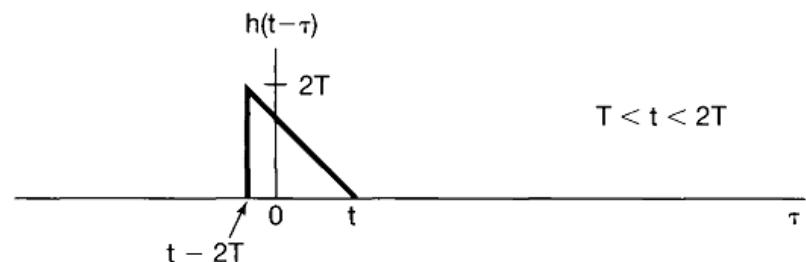
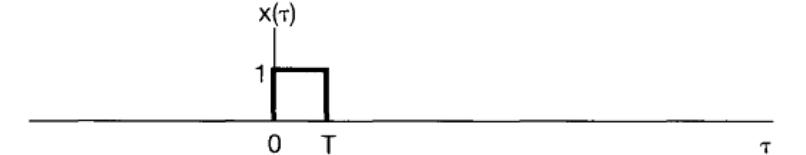
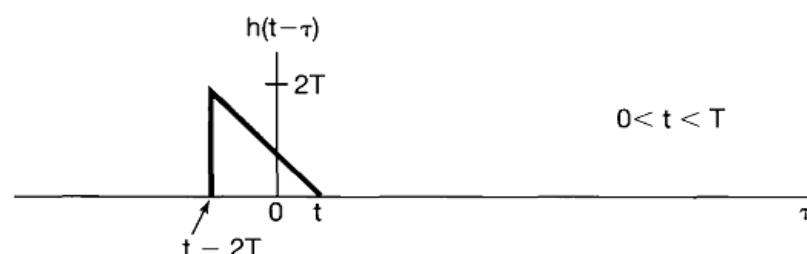
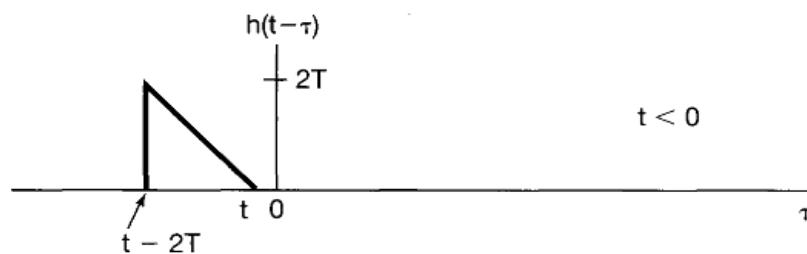
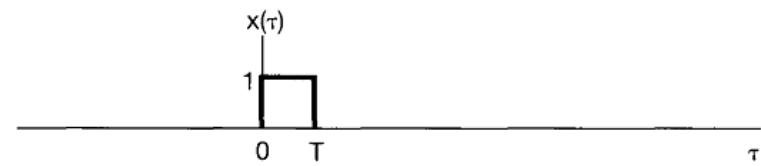
$$= \frac{1}{a} (1 - e^{-at})$$



■ Example 2.7: $y(t) = \int_{-\infty}^{+\infty} x(\tau)h(t - \tau)d\tau$

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

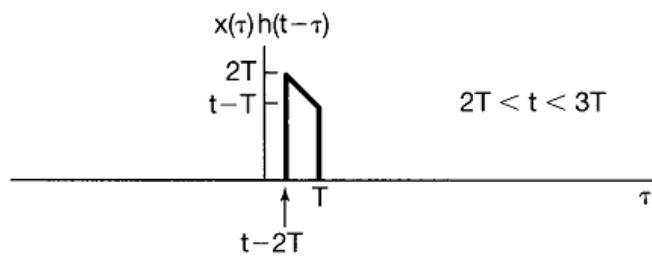
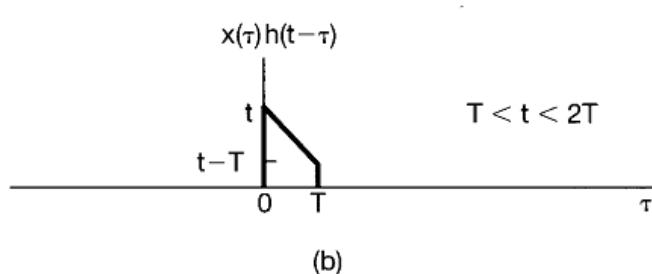
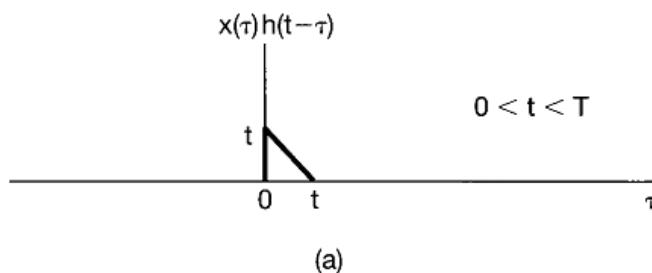
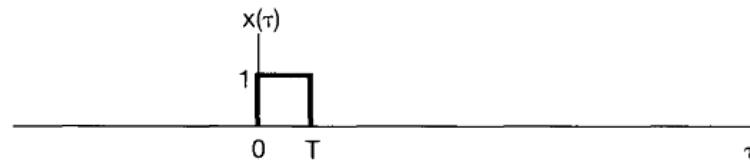
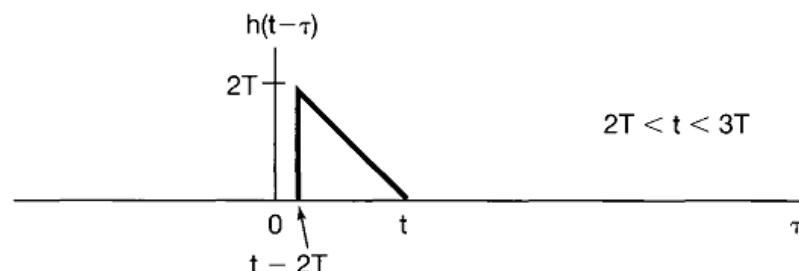
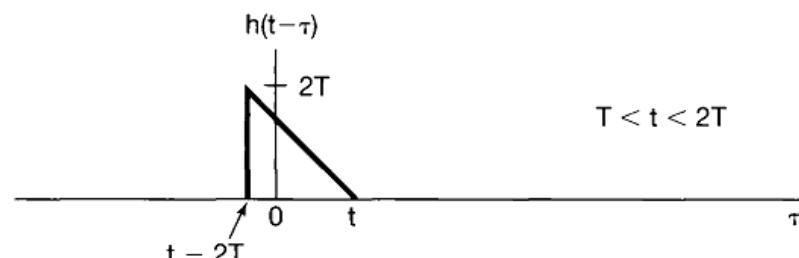
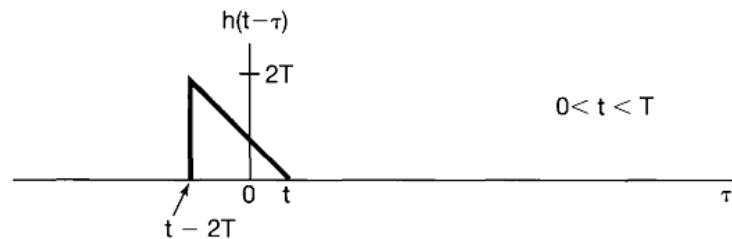
$$h(t) = \begin{cases} t, & 0 < t < 2T \\ 0, & \text{otherwise} \end{cases}$$



■ Example 2.7: $y(t) = \int_{-\infty}^{+\infty} x(\tau)h(t - \tau)d\tau$

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

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

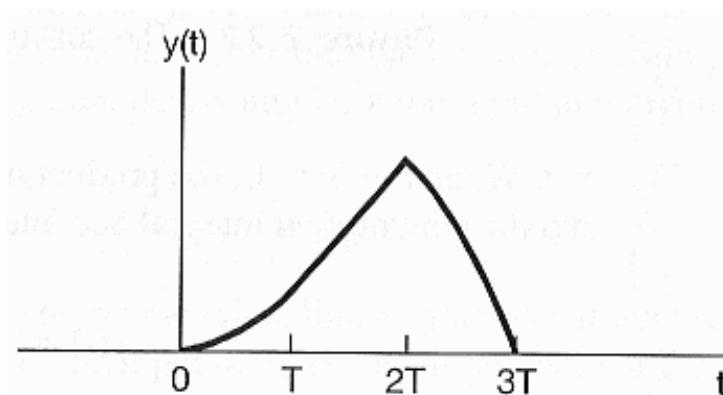


■ Example 2.7:

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

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

$$y(t) = \begin{cases} 0, & t < 0 \\ \frac{1}{2}t^2, & 0 < t < T \\ Tt - \frac{1}{2}T^2, & T < t < 2T \\ -\frac{1}{2}t^2 + Tt + \frac{3}{2}T^2, & 2T < t < 3T \\ 0, & 3T < t \end{cases}$$

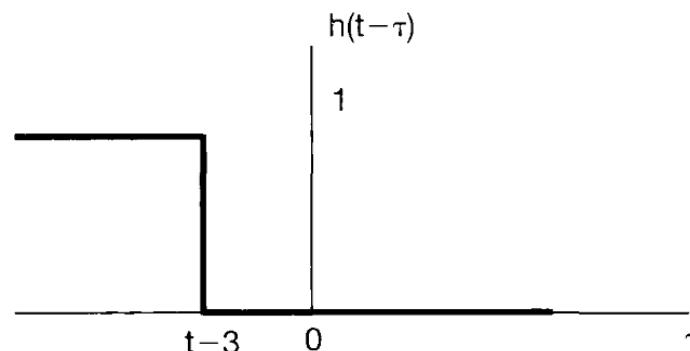
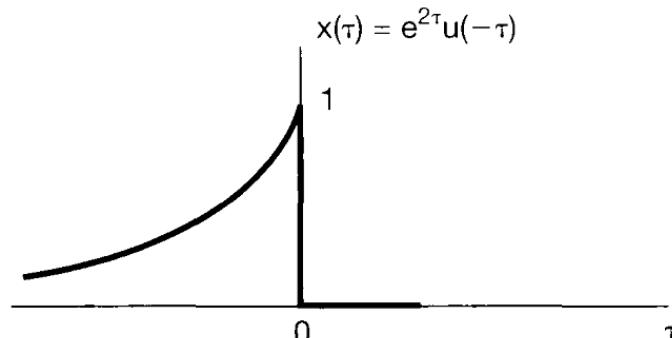


■ Example 2.8:

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

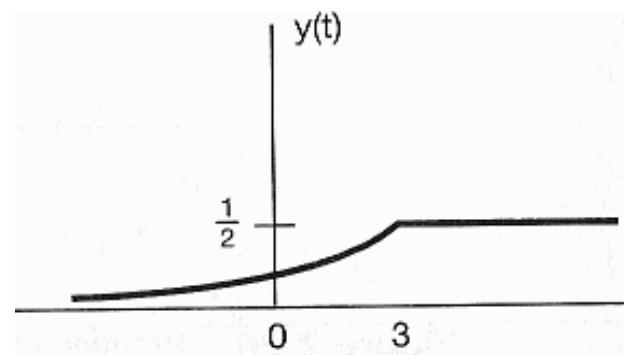
$$x(t) = e^{2t} u(-t)$$

$$h(t) = u(t - 3)$$



$$\begin{aligned} \text{for } t - 3 \leq 0, \quad y(t) &= \int_{-\infty}^{t-3} e^{2\tau} d\tau \\ &= \frac{1}{2}e^{2(t-3)} \end{aligned}$$

$$\begin{aligned} \text{for } t - 3 \geq 0, \quad y(t) &= \int_{-\infty}^0 e^{2\tau} d\tau \\ &= \frac{1}{2} \end{aligned}$$



■ Signal and System:

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

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

$$x(t) \rightarrow \boxed{h(t)} \rightarrow y(t)$$

All the following systems have $h[n]$ as their impulse response

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

$$y[n] = (x[n] + x[n-1])^2,$$

$$y[n] = \max(x[n], x[n-1]).$$

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

Only the third system is LTI and $h[n]$ describes its complete characteristics!

- Discrete-Time Linear Time-Invariant Systems

- The convolution sum

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

- Continuous-Time Linear Time-Invariant Systems

- The convolution integral

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

- Properties of Linear Time-Invariant Systems

- Causal Linear Time-Invariant Systems

Described by Differential & Difference Equations

- Singularity Functions

1. Commutative property

$$\begin{aligned} a \times b &= b \times a \\ a + b &= b + a \end{aligned}$$

2. Distributive property

$$a \times (b + c) = a \times b + a \times c$$

3. Associative property

$$\begin{aligned} a \times (b \times c) &= (a \times b) \times c \\ &= \dots = a \times b \times c \end{aligned}$$

4. With or without memory**5. Invertibility****6. Causality****7. Stability****8. Unit step response**

■ Commutative Property:

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

$$\begin{aligned}
 x(t) * h(t) &= \int_{-\infty}^{+\infty} x(\tau)h(t-\tau)d\tau & t - \tau &= \sigma \\
 &= \int_{+\infty}^{-\infty} x(t-\sigma)h(\sigma)(-d\sigma) & -d\sigma &= d\sigma \\
 &= \int_{-\infty}^{+\infty} x(t-\sigma)h(\sigma)d\sigma \\
 &= \int_{-\infty}^{+\infty} h(\sigma)x(t-\sigma)d\sigma = h(t) * x(t)
 \end{aligned}$$

$$a \times (b + c) = a \times b + a \times c$$

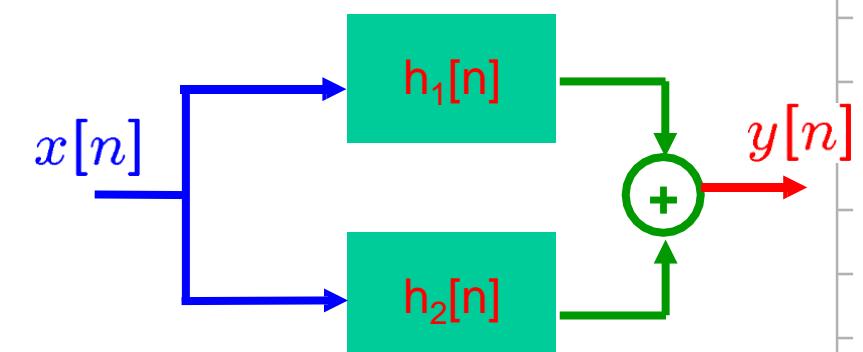
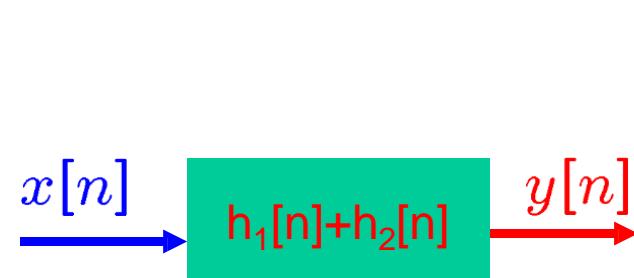
Distributive Property:

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

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

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

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



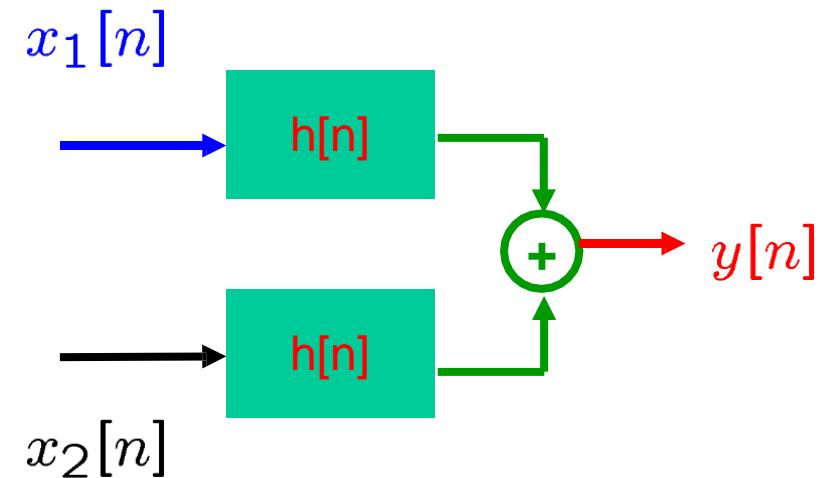
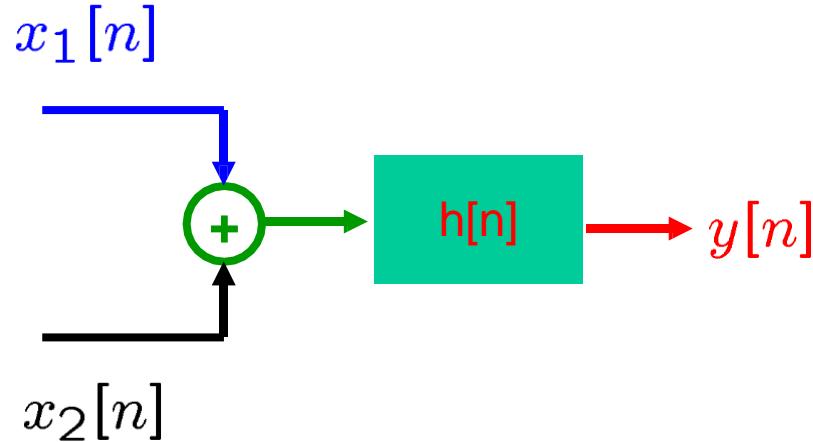
Distributive Property:

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

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

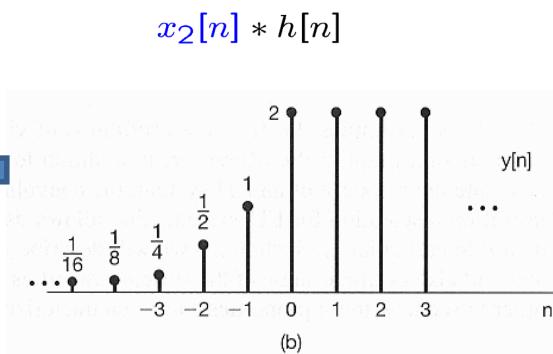
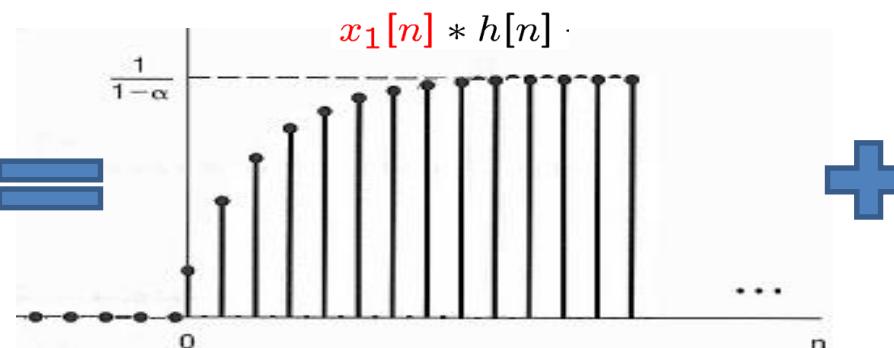
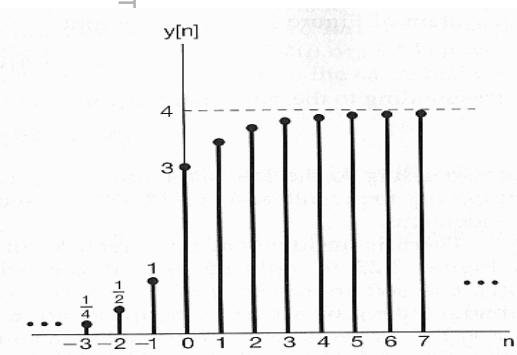
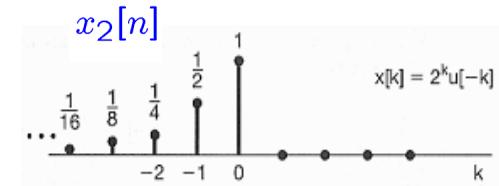
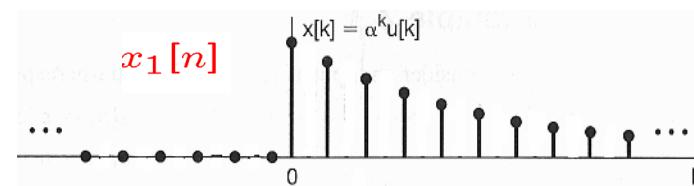
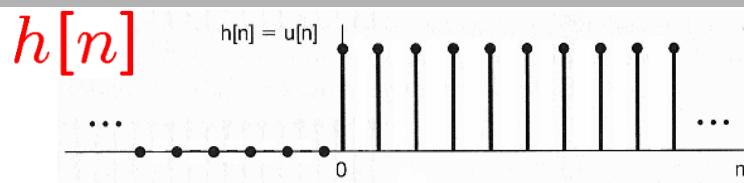
$$(x_1[n] + x_2[n]) * h[n] = x_1[n] * h[n] + x_2[n] * h[n]$$

$$(x_1(t) + x_2(t)) * h(t) = x_1(t) * h(t) + x_2(t) * h(t)$$



- Example 2.10

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



$$\begin{aligned} y[n] &= x[n] * h[n] \\ &= (x_1[n] + x_2[n]) * h[n] = x_1[n] * h[n] + x_2[n] * h[n] \end{aligned}$$

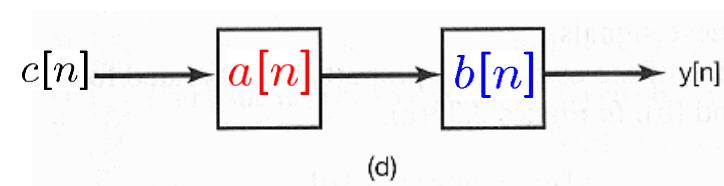
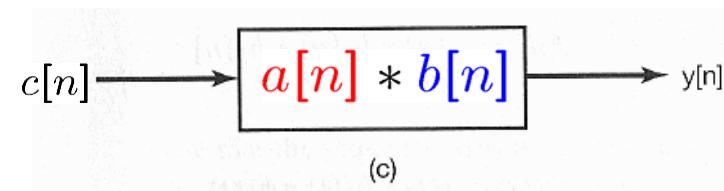
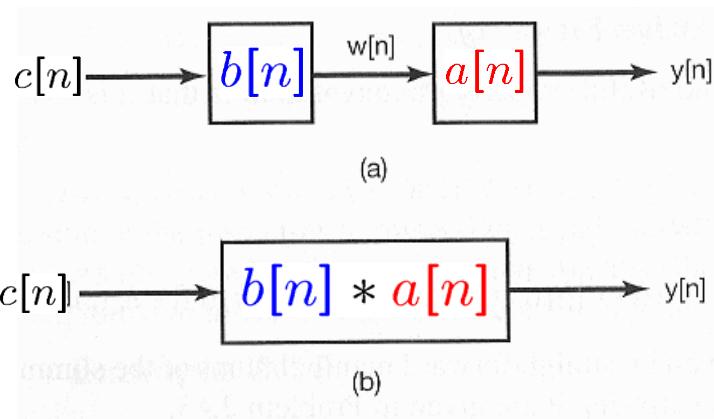
■ Associative Property:

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

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

$$a[n] * (b[n] * c[n]) = (a[n] * b[n]) * c[n]$$

$$a(t) * (b(t) * c(t)) = (a(t) * b(t)) * c(t)$$



■ Systems with or without memory

■ Memoryless systems

- Output depends only on the input **at that same time**

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

$$y(t) = Rx(t) \quad (\text{resistor})$$

■ Systems with memory

$$y[n] = \sum_{k=-\infty}^n x[k] \quad (\text{accumulator})$$

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

$$y[n] = x[n - 1] \quad (\text{delay})$$

■ Memoryless:

- A DT LTI system is **memoryless** if

$$x[n] * h[n] = \sum_{k=-\infty}^{+\infty} x[k]h[n-k]$$
$$h[n] = 0 \text{ for } n \neq 0$$

- The **impulse response**:

$$h[n] = K\delta[n], \quad K = h[0]$$

- The **convolution sum**:

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

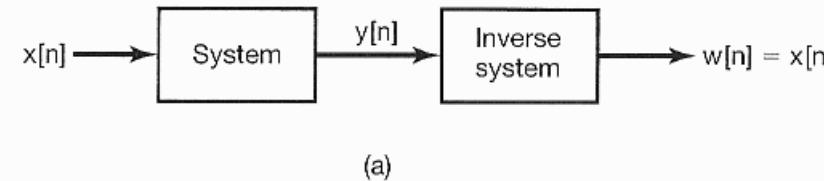
- Similarly, for CT LTI system:

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

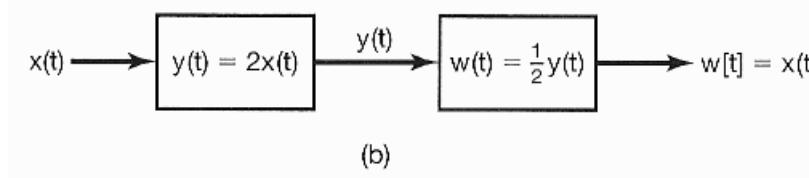
▪ Invertibility & Inverse Systems

▪ Invertible systems

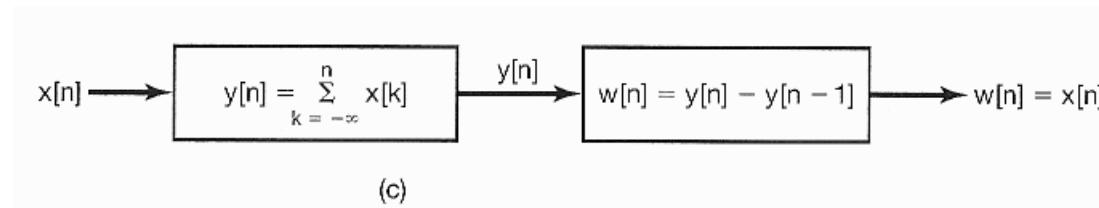
- Distinct inputs lead to distinct outputs



(a)



(b)



(c)

$y(t) = x(t)^2$ is not invertible

■ Invertibility:

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

$$x(t) \rightarrow \boxed{h_1(t)} \rightarrow y(t) \rightarrow \boxed{h_2(t)} \rightarrow w(t)$$

$$y(t) = x(t) * h_1(t) \quad w(t) = y(t) * h_2(t)$$

$$\Rightarrow w(t) = x(t) * h_1(t) * h_2(t)$$

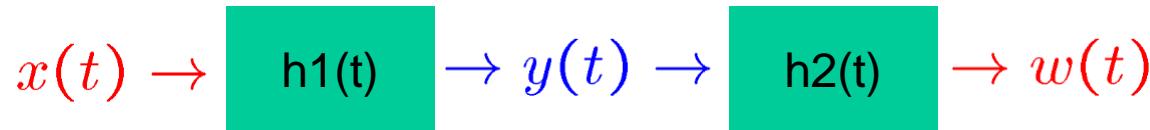
$$x(t) \rightarrow \boxed{\text{Identity System } \delta(t)} \rightarrow x(t)$$

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

$$\implies h_2(t) * h_1(t) = \delta(t)$$

■ Example 2.11: Pure time shift

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



- $y(t) = x(t - t_0)$
 - delay if $t_0 > 0$
 - advance if $t_0 < 0$

$$\Rightarrow h_1(t) = \delta(t - t_0)$$

$$\Rightarrow \boxed{x(t) * \delta(t - t_0) = x(t - t_0)}$$

- $w(t) = x(t) = y(t + t_0)$

$$\Rightarrow h_2(t) = \delta(t + t_0) \Rightarrow y(t) * \delta(t + t_0) = y(t + t_0)$$

$$\Rightarrow h_1(t) * h_2(t) = \delta(t - t_0) * \delta(t + t_0) = \delta(t)$$

- Example 2.12

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



$$h_1[n] = u[n]$$

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

\Rightarrow a **running-sum** operation

- Its inverse is a **first difference** operation:

$$w[n] = y[n] - y[n-1] \Rightarrow h_2[n] = \delta[n] - \delta[n-1]$$

$$\Rightarrow h_1[n] * h_2[n] = u[n] - u[n-1] = \delta[n]$$

■ Causality:

- The **output** of a **causal** system depends only on the **present and past** values of the **input** to the system
- Specifically, $y[n]$ must **not** depend on $x[k]$, for $k > n$

$$h[n - k] = 0, \quad \text{for } k > n$$

$$h[n] = 0, \quad \text{for } n < 0$$

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

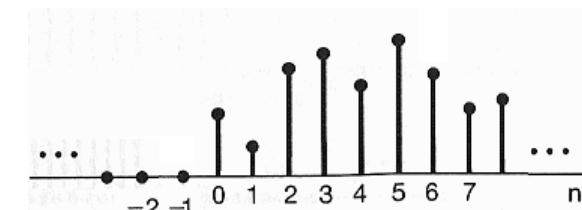
$$x(t) * h(t) = \int_{-\infty}^{+\infty} x(\tau)h(t - \tau)d\tau$$

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

- It implies that the system is **initially rest**

- A **CT** LTI system is **causal** if

$$h(t) = 0, \quad \text{for } t < 0$$



▪ Convolution Sum & Integral

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

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

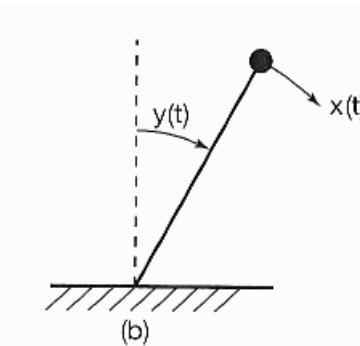
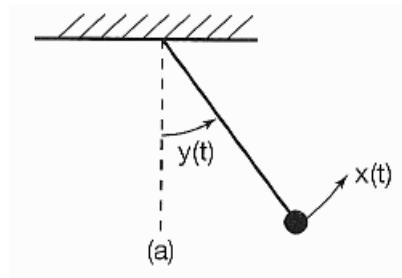
$$= \sum_{k=0}^{\infty} h[k] x[n-k]$$

$$\begin{aligned} y(t) &= \int_{-\infty}^{\infty} x(\tau) h(t-\tau) d\tau = \int_0^{\infty} h(\tau) x(t-\tau) d\tau \\ &= \int_{-\infty}^t x(\tau) h(t-\tau) d\tau \end{aligned}$$

■ Stability

■ Stable systems

- Small inputs lead to responses that **do not diverge**
- Every bounded input excites a **bounded output**
 - Bounded-input bounded-output stable (**BIBO stable**)
 - For all $|x(t)| < a$, then $|y(t)| < b$, for all t



- Balance in a bank account?

$$y[n] = 1.01y[n - 1] + x[n]$$

■ **Stability:**

- A system is **stable** if every **bounded input** produces a **bounded output**

$$x[n] \rightarrow \text{Stable LTI} \rightarrow y[n]$$

$$|x[n]| < B \quad \text{for all } n$$

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

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

$$\Rightarrow |y[n]| \leq B \left(\sum_{k=-\infty}^{+\infty} |h[k]| \right)$$

if $\sum_{k=-\infty}^{+\infty} |h[k]| < \infty$, i.e., **absolutely summable** then, $y[n]$ is **bounded**

■ Stability:

- For CT LTI stable system:

$$x(t) \rightarrow \text{Stable LTI} \rightarrow y(t)$$

$$|x(t)| < B \quad \text{for all } t \quad |y(t)| = \left| \int_{-\infty}^{+\infty} h(\tau) x(t - \tau) d\tau \right|$$

$$\Rightarrow |y(t)| \leq \int_{-\infty}^{+\infty} |h(\tau)| |x(t - \tau)| d\tau$$

$$\Rightarrow |y(t)| \leq B \left(\int_{-\infty}^{+\infty} |h(\tau)| d\tau \right)$$

if $\int_{-\infty}^{+\infty} |h(\tau)| d\tau < \infty$, i.e., absolutely integrable then, $y(t)$ is bounded

- Example 2.13: Pure time shift

- $y[n] = x[n - n_0]$ & $h[n] = \delta[n - n_0]$

- $y(t) = x(t - t_0)$ & $h(t) = \delta(t - t_0)$

$$\Rightarrow \sum_{n=-\infty}^{+\infty} |h[n]| = \sum_{n=-\infty}^{+\infty} |\delta[n - n_0]| = 1 \quad \text{absolutely summable}$$

$$\Rightarrow \int_{-\infty}^{+\infty} |h(\tau)| = \int_{-\infty}^{+\infty} |\delta(\tau - t_0)| d\tau = 1 \quad \text{absolutely integrable}$$

\Rightarrow A (CT or DT) pure time shift is **stable**

- Example 2.13: Accumulator

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

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

$$\Rightarrow \sum_{n=-\infty}^{+\infty} |h[n]| = \sum_{n=0}^{+\infty} |u[n]| = \infty \quad \text{NOT absolutely summable}$$

$$\Rightarrow \int_{-\infty}^{+\infty} |h(\tau)| = \int_0^{\infty} |u(\tau)| d\tau = \infty \quad \text{NOT absolutely integrable}$$

\Rightarrow A accumulator or integrator is NOT stable

■ Discrete-Time Linear Time-Invariant Systems

- The convolution sum

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

■ Continuous-Time Linear Time-Invariant Systems

- The convolution integral

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

■ Properties of Linear Time-Invariant Systems

1. Commutative property
2. Distributive property
3. Associative property
4. With or without memory
5. Invertibility
6. Causality
7. Stability
8. Unit step response

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

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

$$a(t) * (b(t) * c(t)) = (a(t) * b(t)) * c(t)$$

$$h(t) = 0 \text{ for } t \neq 0 \quad h(t) = 0, \quad \text{for } t < 0$$

$$h_2(t) * h_1(t) = \delta(t) \quad \text{if } \int_{-\infty}^{+\infty} |h(\tau)|d\tau < \infty$$

■ Causal Linear Time-Invariant Systems

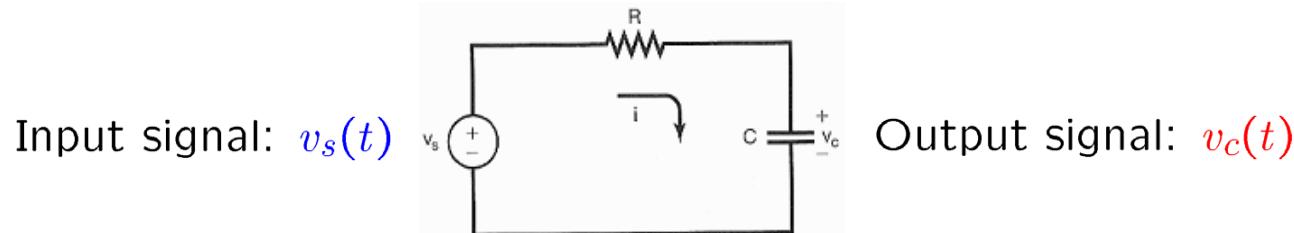
Described by Differential & Difference Equations

■ Singularity Functions

$$u_k(t) * u_r(t) = u_{k+r}(t)$$

■ Linear Constant-Coefficient Differential Equations

- Describe an important class of continuous systems.
- In **Diff Eqn**, we have learned a way to analyze their solutions.
- In subsequent chapters, additional tools for the analysis of systems described by such equations are developed.
- Provide an **implicit specification** of the system



$$\Rightarrow \frac{dv_c(t)}{dt} + \frac{1}{RC}v_c(t) = \frac{1}{RC}v_s(t)$$

$$x(t) \rightarrow \text{RC Circuit} \rightarrow y(t) \Rightarrow \frac{d}{dt}y(t) + a y(t) = b x(t)$$

Example 2.14

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

$$y(0) = 0$$

$$x(t) = Ke^{3t}u(t) \Rightarrow y(t) = ?$$

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

$y_h(t)$ solution of homogeneous equation, $x(t) = 0$

$y_p(t)$ particular response, $x(t) = Ke^{3t}u(t)$

Example 2.14

homogeneous response

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

$$y(0) = 0$$

$$x(t) = Ke^{3t}u(t) \implies y(t) = ?$$

$$\frac{dy_h(t)}{dt} + 2y_h(t) = 0$$

$$y_h(t) = Ae^{st} \implies sAe^{st} + 2Ae^{st} = 0$$

$$\implies (s + 2)Ae^{st} = 0 \implies s = -2$$

$$y_h(t) = Ae^{-2t}$$

Example 2.14

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

$$y(0) = 0$$

$$x(t) = Ke^{3t}u(t) \Rightarrow y(t) = ?$$

Particular response for $t > 0$

$$\frac{dy_p(t)}{dt} + 2y_p(t) = Ke^{3t} \quad t \geq 0$$

$$y_p(t) = Be^{3t} \Rightarrow 3Be^{3t} + 2Be^{3t} = Ke^{3t} \Rightarrow B = K/5$$

$$y_p(t) = Ke^{3t}/5$$

$$y(t) = Ae^{-2t} + Ke^{3t}/5$$

$$y(0) = A + K/5 = 0 \Rightarrow A = -K/5$$

Example 2.14

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

$$y(0) = 0$$

$$x(t) = Ke^{3t}u(t) \Rightarrow y(t) = ?$$

Particular response for $t < 0$

$$\frac{dy_p(t)}{dt} + 2y_p(t) = 0 \quad t < 0$$

$$y_p(t) = 0$$

$$y(t) = Ae^{-2t}$$

$$y(0) = 0 = A$$

■ Linear Constant-Coefficient Differential Equations

- The response of a linear system to zero input is zero !

$$x_1(t) = 0 \implies y_1(t) = a$$

$$x_2(t) = Kx_1(t) = 0 \implies y_2 = Ky_1(t) = Ka \quad K \neq 1$$

$$Ka = a \rightarrow a = 0.$$

- So, when the auxiliary condition is not zero, the system is not linear

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

$$y(0) = \alpha \neq 0$$

■ Linear Constant-Coefficient Differential Equations

- For a causal system, we have

$$x_1(t) = x_2(t) \quad \forall t < t_0$$

$$\implies$$

$$y_1(t) = y_2 \quad \forall t < t_0$$

- So, a causal linear system should be initially at rest!

$$x_1(t) = 0 \quad \forall t \implies y_1(t) = 0 \quad \forall t$$

$$x_2(t) = 0 \quad \forall t < t_0 \implies y_2 = 0 \quad \forall t < t_0$$

- Linear Constant-Coefficient Differential Equations

- For a general CT LTI system, with N-th order,

$$x(t) \rightarrow \text{CT LTI} \rightarrow y(t)$$

$$a_N \frac{d^N}{dt^N} y(t) + a_{N-1} \frac{d^{N-1}}{dt^{N-1}} y(t) + \cdots + a_1 \frac{d}{dt} y(t) + a_0 y(t)$$

$$= b_M \frac{d^M}{dt^M} x(t) + b_{M-1} \frac{d^{M-1}}{dt^{M-1}} x(t) + \cdots + b_1 \frac{d}{dt} x(t) + b_0 x(t)$$

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

To completely determine the answer $y(t_0), \dots, y^{N-1}(t_0)$, should be given

■ Linear Constant-Coefficient Differential Equations

- The linear ODEs without specifying initial conditions, do not declare a unique system.
- Consider

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

If $y(t_0), \dots, y^{N-1}(t_0)$ Is not declared,  No specific system is implied

If initial condition $y(t_0), \dots, y^{N-1}(t_0) \neq 0$  The system is not linear

If t_0 in $y(t_0)$ Is fixed  The system is neither causal nor time invariant

- The linear Constant Coefficient Differential Equation Which is initially at rest specifies a causal LTI system

■ Linear Constant-Coefficient Differential Equations

- The linear constant coefficient difference equation without specifying initial conditions, does not declare a unique system.
- Consider

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

If $y[n_0 - 1], \dots, y[n_0 - N]$ Is not declared,  No specific system is implied

If initial condition $y[n_0 - 1], \dots, y[n_0 - N] \neq 0$  The system is not linear

If n_0 in $y[n_0]$ Is fixed  The system is neither causal nor time invariant

- The linear Constant Coefficient Difference Equation Which is initially at rest specifies a causal LTI system

- Linear Constant-Coefficient Difference Equations

- For a general DT LTI system, with N-th order,



$$a_0 y[n] + a_1 y[n-1] + \cdots + a_{N-1} y[n-N+1] + a_N y[n-N]$$

$$= b_0 x[n] + b_1 x[n-1] + \cdots + b_{M-1} x[n-M+1] + b_M x[n-M]$$

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

$$\Rightarrow h[n] = ?$$

- Recursive Equation:

$$a_0 y[n] + a_1 y[n - 1] + \cdots + a_{N-1} y[n - N + 1] + a_N y[n - N]$$

$$= b_0 x[n] + b_1 x[n - 1] + \cdots + b_{M-1} x[n - M + 1] + b_M x[n - M]$$

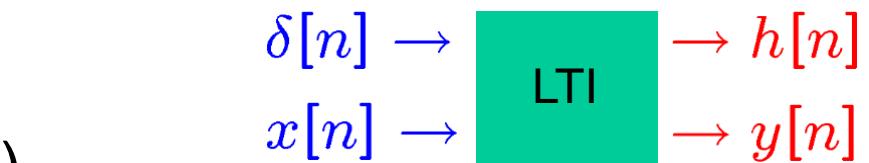
$$\sum_{k=0}^N a_k y[n - k] = \sum_{k=0}^M b_k x[n - k]$$

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

- Recursive Equation:

- For example, (Example 2.15)

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

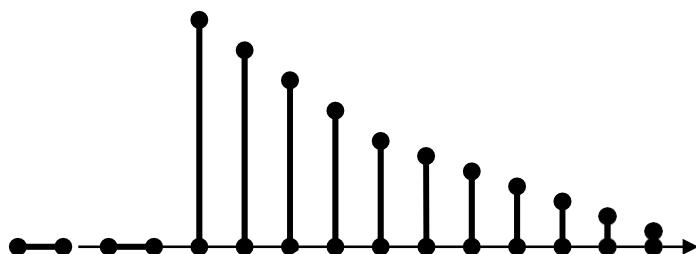


$$y[n] = 0, \quad \text{for } n \leq -1$$

$$x[n] = K \delta[n]$$

$$\Rightarrow \begin{cases} y[0] = x[0] + \frac{1}{2}y[-1] & = K \\ y[1] = x[1] + \frac{1}{2}y[0] & = K \cdot \frac{1}{2} \\ y[2] = x[2] + \frac{1}{2}y[1] & = K \cdot \left(\frac{1}{2}\right)^2 \\ \vdots & \\ y[n] = x[n] + \frac{1}{2}y[n-1] & = K \cdot \left(\frac{1}{2}\right)^n \end{cases}$$

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



\Rightarrow an Infinite Impulse Response (IIR) system

Nonrecursive Equation:

- When $N = 0$,

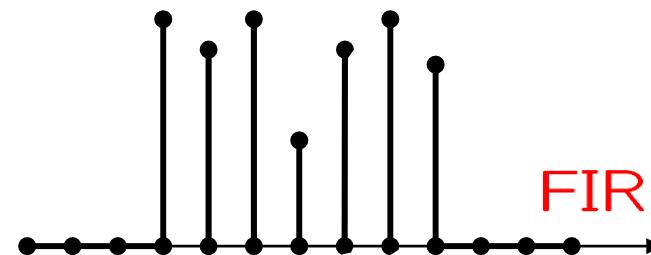
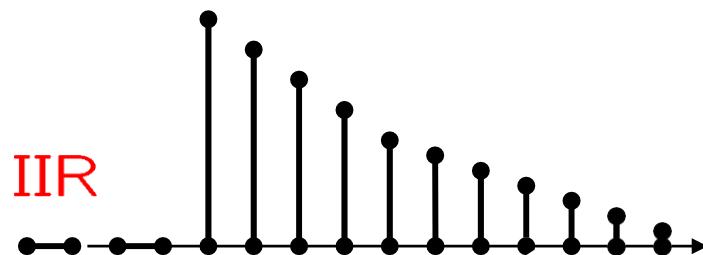
$$\sum_{k=0}^N a_k y[n-k] = \sum_{k=0}^M b_k x[n-k]$$

$$\Rightarrow y[n] = \sum_{k=0}^M \left(\frac{b_k}{a_0} \right) x[n-k]$$

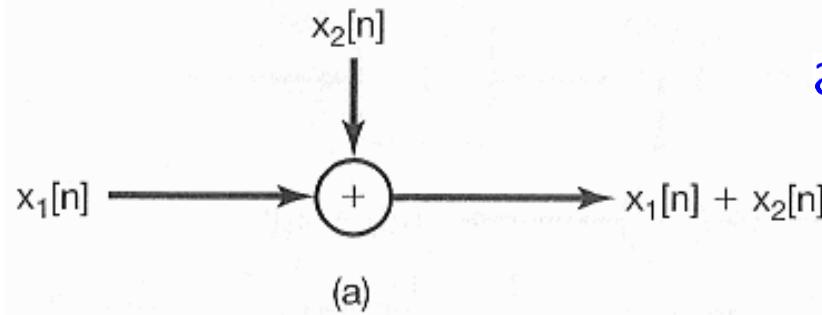
$$y[n] = \sum_{k=-\infty}^{+\infty} h[k] x[n-k]$$

$$\Rightarrow h[n] = \begin{cases} \frac{b_n}{a_0}, & 0 \leq n \leq M \\ 0, & \text{otherwise} \end{cases}$$

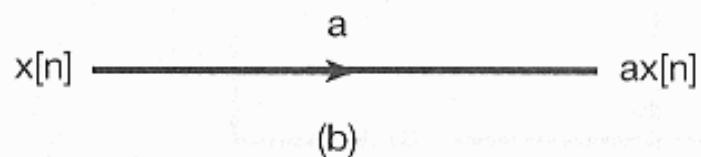
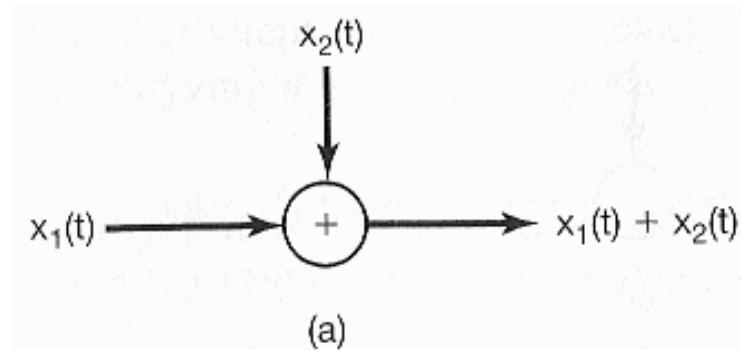
⇒ a Finite Impulse Response (FIR) system



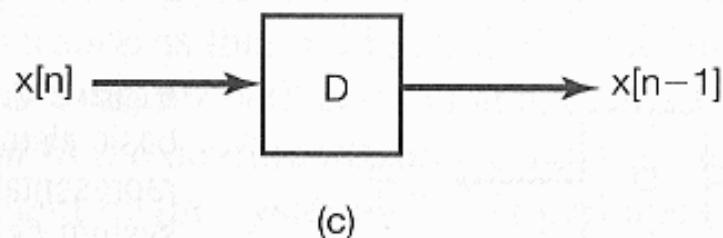
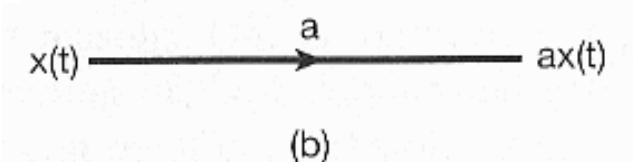
- Block Diagram Representations:



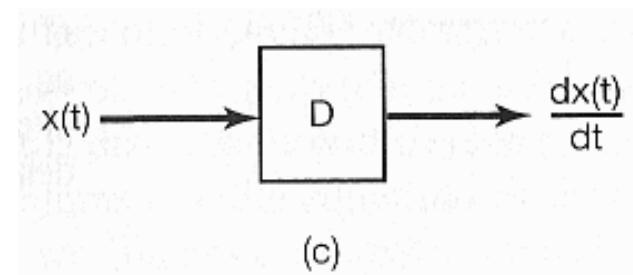
an adder



multiplication
by a coefficient



a unit delay/
differentiator



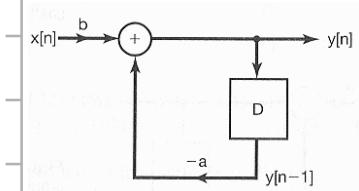
- Block Diagram Representations:

$$y[n] + ay[n-1] = bx[n]$$

$$\frac{d}{dt}y(t) + ay(t) = bx(t)$$

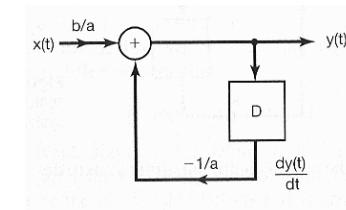
$$y[n] = -ay[n-1] + bx[n]$$

$$y(t) = -\frac{1}{a}\frac{d}{dt}y(t) + \frac{b}{a}x(t)$$



$$D \iff z^{-1}$$

$$D \iff s$$



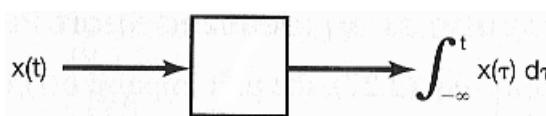
■ Block Diagram Representations:

- In practice, implementation of derivative block is difficult and it would be sensitive to noise.
- So, alternatively, we can use integral block

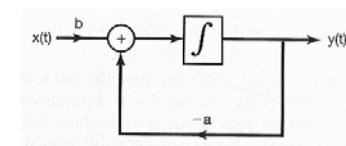
$$\frac{d}{dt}y(t) = bx(t) - ay(t)$$

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

$$\Rightarrow y(t) = y(t_0) + \int_{t_0}^t [bx(\tau) - ay(\tau)] d\tau$$



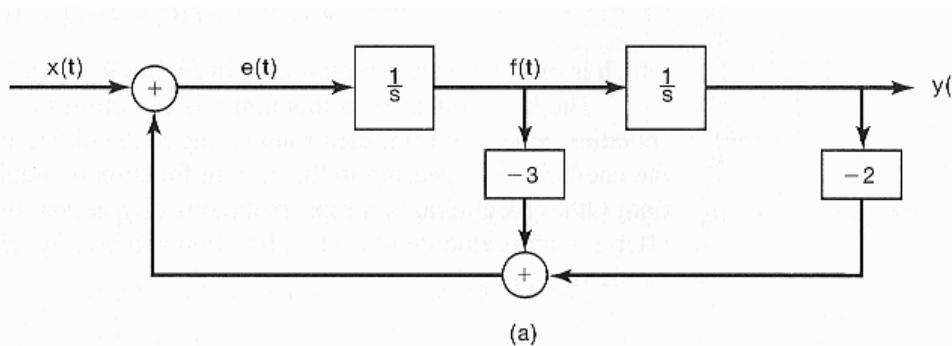
$$\int \iff \frac{1}{s}$$



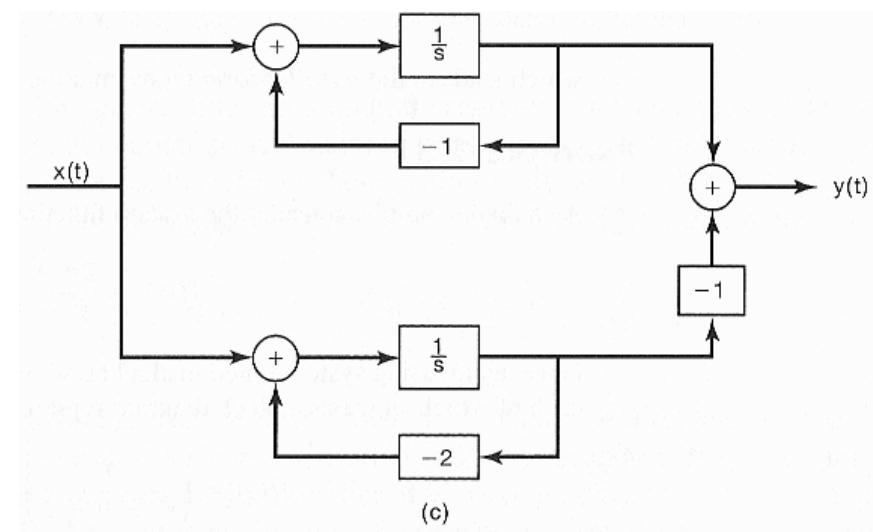
■ Block Diagram Representations:

$$\frac{d^2}{dt^2}y(t) + 3\frac{d}{dt}y(t) + 2y(t) = x(t)$$

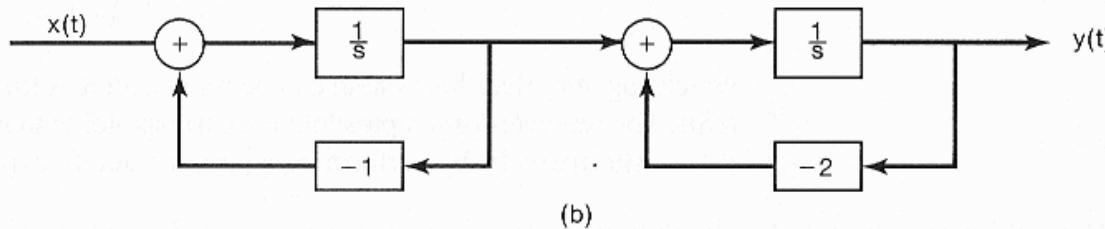
$$\int \iff \frac{1}{s}$$



(a)



(c)



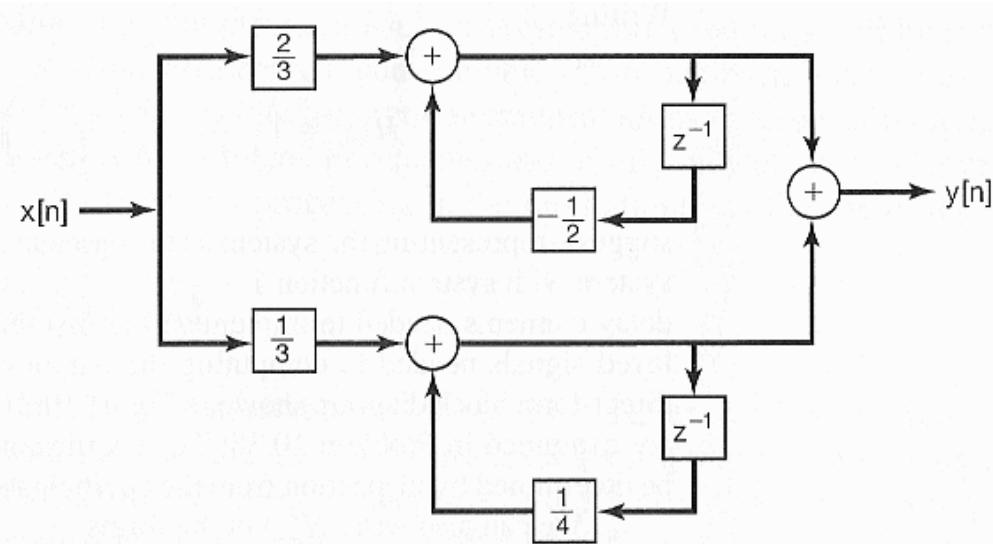
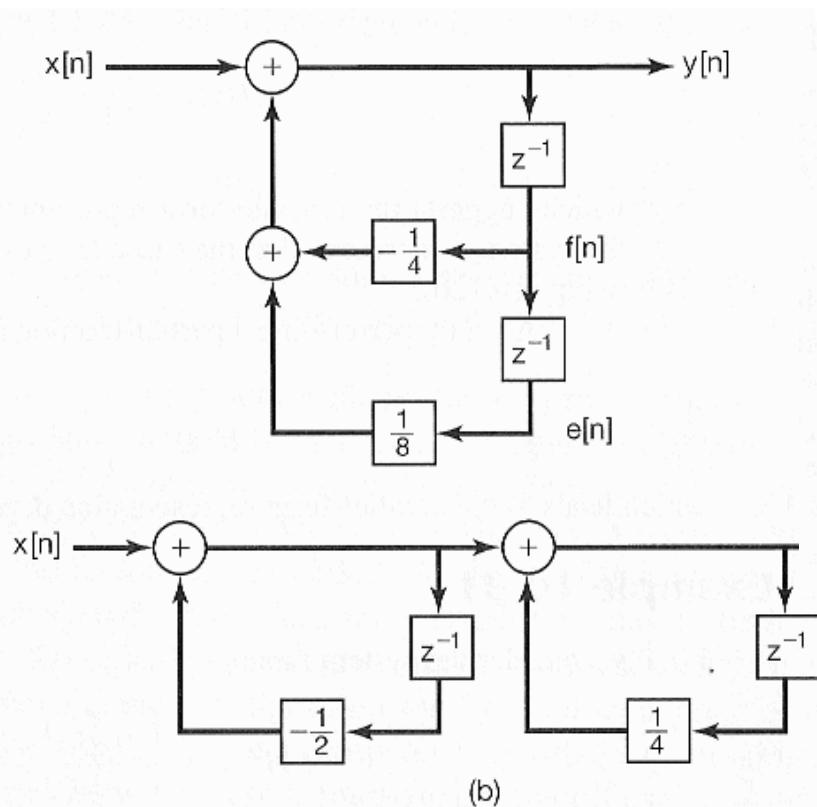
(b)

- Example 9.30 (pp.711)

$$H(s) = \frac{1}{(s+1)(s+2)}$$

- Block Diagram Representations:

$$y[n] + \frac{1}{4}y[n-1] - \frac{1}{8}y[n-2] = x[n]$$



- Example 10.30 (pp.786)

$$H(z) = \frac{1}{(1 + \frac{1}{2}z^{-1})(1 - \frac{1}{4}z^{-1})}$$

■ Discrete-Time Linear Time-Invariant Systems

- The convolution sum

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

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

■ Continuous-Time Linear Time-Invariant Systems

- The convolution integral

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

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

■ Properties of Linear Time-Invariant Systems

1. Commutative property
2. Distributive property
3. Associative property
4. With or without memory
5. Invertibility
6. Causality
7. Stability
8. Unit step response

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

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

$$a(t) * (b(t) * c(t)) = (a(t) * b(t)) * c(t)$$

$$h[n] = 0 \text{ for } n \neq 0 \quad h(t) = 0, \quad \text{for } t < 0$$

$$h_2(t) * h_1(t) = \delta(t) \quad \text{if } \int_{-\infty}^{+\infty} |h(\tau)| d\tau < \infty$$

■ Causal Linear Time-Invariant Systems

Described by Differential & Difference Equations

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

■ Singularity Functions

$$u_k(t) * u_r(t) = u_{k+r}(t)$$

■ Unit Step Response:

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

- For an LTI system, its **impulse response** is:

$$\delta[n] \rightarrow \boxed{\text{DT LTI}} \rightarrow h[n]$$

$$\delta(t) \rightarrow \boxed{\text{CT LTI}} \rightarrow h(t)$$

- Its **unit step response** is:

$$u[n] \rightarrow \boxed{\text{DT LTI}} \rightarrow s[n]$$

$$\Rightarrow s[n] = u[n] * h[n]$$

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

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

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

$$u(t) \rightarrow \boxed{\text{CT LTI}} \rightarrow s(t)$$

$$\Rightarrow s(t) = u(t) * h(t)$$

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

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

$$\Rightarrow h(t) = \frac{ds(t)}{dt}$$

■ Discrete-Time Linear Time-Invariant Systems

- The convolution sum

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

■ Continuous-Time Linear Time-Invariant Systems

- The convolution integral

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

■ Properties of Linear Time-Invariant Systems

- Commutative property
- Distributive property
- Associative property
- With or without memory
- Invertibility
- Causality
- Stability
- Unit step response

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

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

$$a(t) * (b(t) * c(t)) = (a(t) * b(t)) * c(t)$$

$$h[n] = 0 \text{ for } n \neq 0 \quad h(t) = 0, \quad \text{for } t < 0$$

$$h_2(t) * h_1(t) = \delta(t) \quad \text{if } \int_{-\infty}^{+\infty} |h(\tau)|d\tau < \infty$$

■ Causal Linear Time-Invariant Systems

Described by Differential & Difference Equations

■ Singularity Functions

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

■ Singularity Functions

- CT unit impulse function is one of **singularity** functions

$$\delta(t) = u_0(t)$$

$$u(t) = u_{-1}(t)$$

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

$$\frac{d}{dt} \delta(t) = u_1(t)$$

$$\int_{-\infty}^t u(\tau) d\tau = u_{-2}(t)$$

$$\frac{d^2}{dt^2} \delta(t) = u_2(t)$$

$$\int_{-\infty}^t \left(\int_{-\infty}^{\tau} u(\sigma) d\sigma \right) d\tau = u_{-3}(t)$$

$$\frac{d^k}{dt^k} \delta(t) = u_k(t)$$

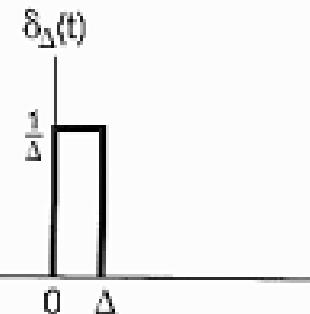
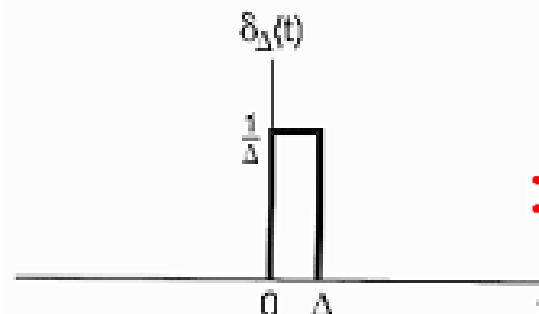
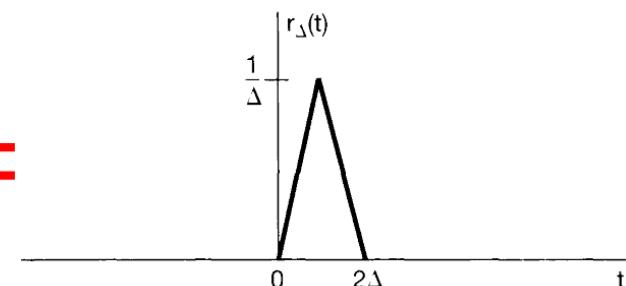
$$\int_{-\infty}^t \cdots \left(\int_{-\infty}^{\tau} u(\sigma) d\sigma \right) \cdots d\tau = u_{-k}(t)$$

▪ Singularity Functions

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

$$\delta(t) = \delta(t) * \delta(t)$$

$$r_{\Delta}(t) = \delta_{\Delta}(t) * \delta_{\Delta}(t)$$

 $*$  $=$ 

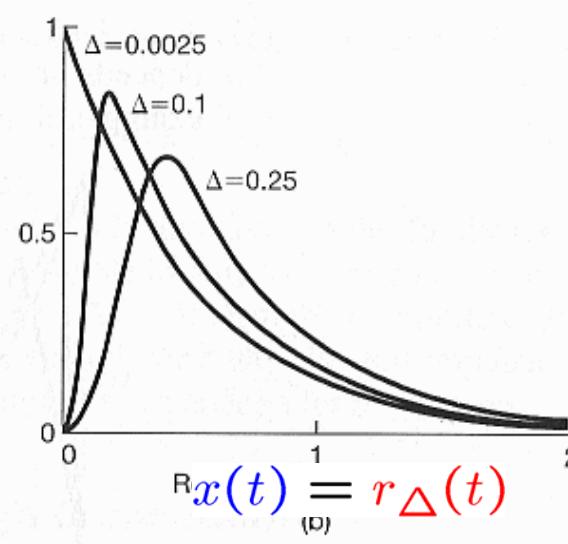
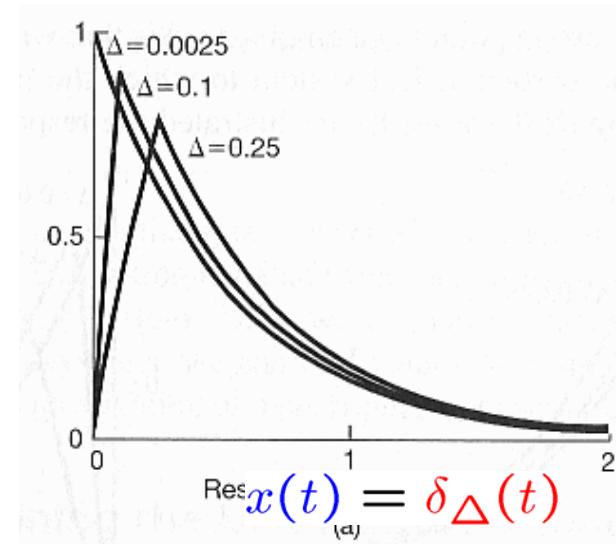
$$\lim_{\Delta \rightarrow 0} \delta_{\Delta}(t) = \delta(t)$$

$$\Rightarrow \lim_{\Delta \rightarrow 0} r_{\Delta}(t) = \delta(t)$$

■ Example 2.16

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

with initial-rest condition



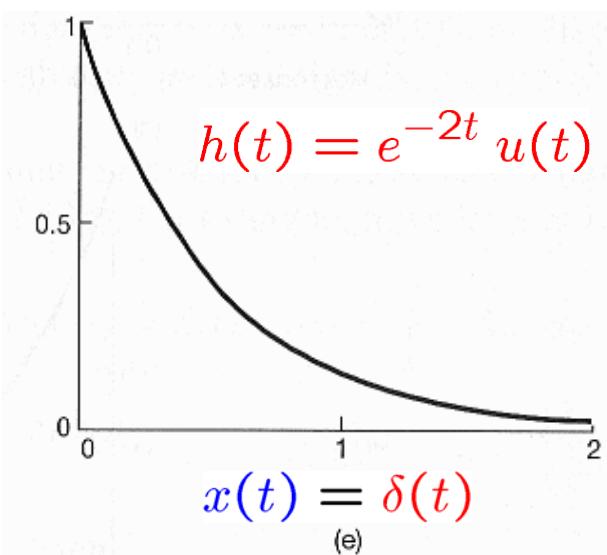
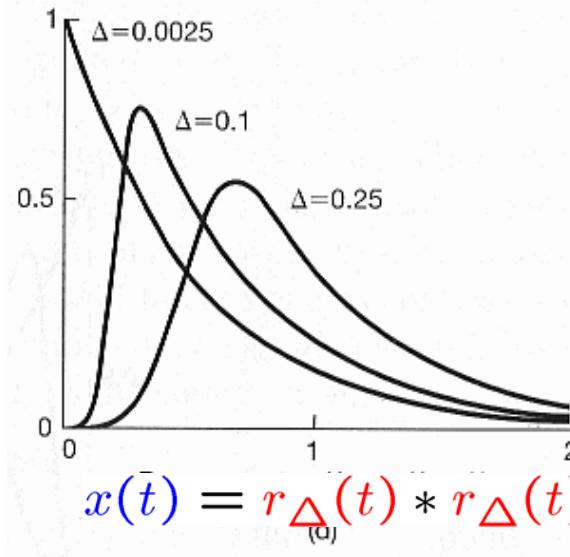
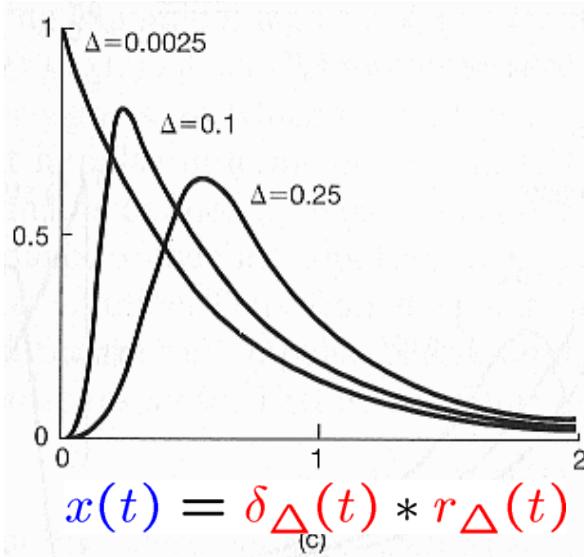
$$x(t) = \delta_{\Delta}(t)$$

$$x(t) = r_{\Delta}(t)$$

$$x(t) = \delta_{\Delta}(t) * r_{\Delta}(t)$$

$$x(t) = r_{\Delta}(t) * r_{\Delta}(t)$$

$$x(t) = \delta(t)$$

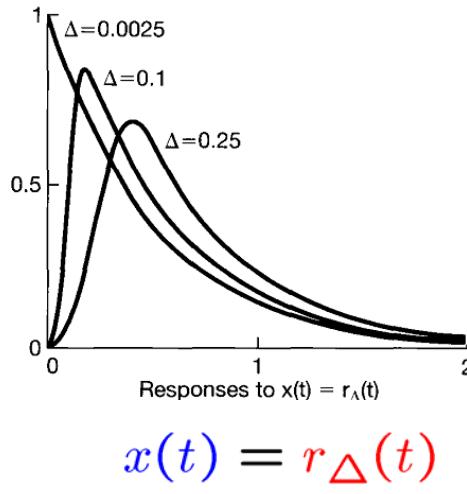
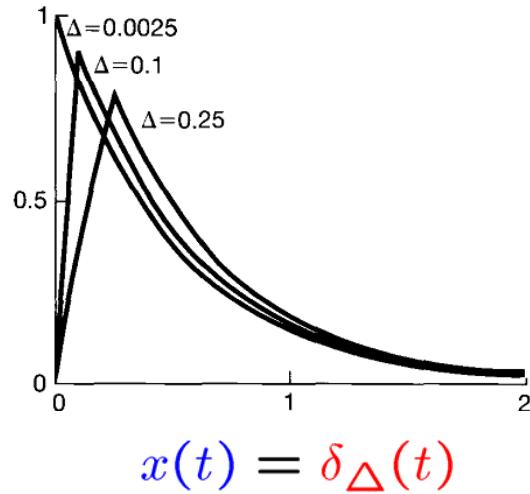


$$h(t) = e^{-2t} u(t)$$

■ Example 2.16

$$\frac{d}{dt}y(t) + 20 y(t) = x(t)$$

with initial-rest condition



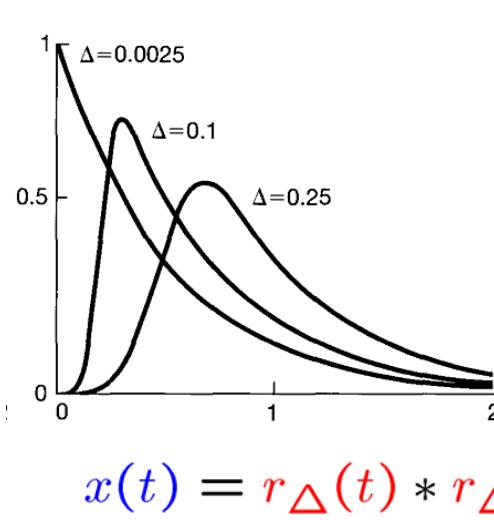
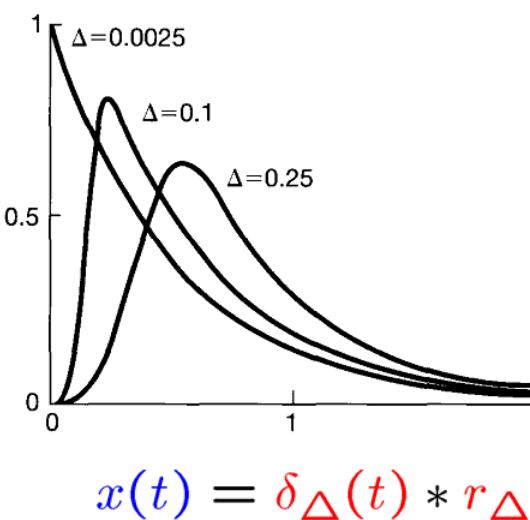
$$x(t) = \delta_{\Delta}(t)$$

$$x(t) = r_{\Delta}(t)$$

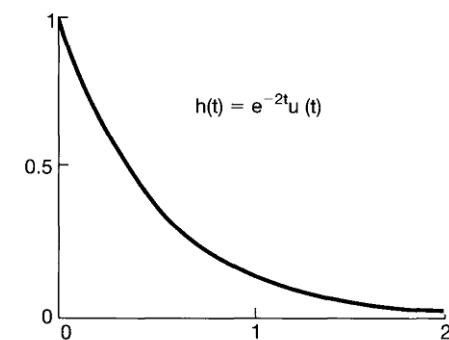
$$x(t) = \delta_{\Delta}(t) * r_{\Delta}(t)$$

$$x(t) = r_{\Delta}(t) * r_{\Delta}(t)$$

$$x(t) = \delta(t)$$



$$h(t) = e^{-20t} u(t)$$



$$x(t) = \delta(t)$$

- Defining the Unit Impulse through Convolution:

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

- Let $x(t) = 1$,

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

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

- So that the unit impulse has unit area

- Unit Doublets of Derivative Operation:

- A system: Output is the **derivative** of input

$$y(t) = \frac{d}{dt}x(t)$$

⇒ The unit impulse response of the system
is the derivative of the unit impulse,
which is called the **unit doublet** $u_1(t)$

- That is, from $x(t) = x(t) * \delta(t)$, we have

$$\frac{d}{dt}x(t) = x(t) * u_1(t)$$

- Unit Doublets of Derivative Operation:

- Similarly,

$$\frac{d^2}{dt^2}x(t) = x(t) * u_2(t)$$

- But,

$$\frac{d^2}{dt^2}x(t) = \frac{d}{dt} \left(\frac{d}{dt}x(t) \right) = (x(t) * u_1(t)) * u_1(t)$$

- Therefore,

$$u_2(t) = u_1(t) * u_1(t)$$

- In general,

$u_k(t)$, $k > 0$, the k th derivative of $\delta(t)$

$$u_k(t) = u_1(t) * \cdots * u_1(t)$$

- Unit Doublets of Integration Operation:

- A system: Output is the **integral** of input

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

- Therefore,

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

- Hence, we have

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

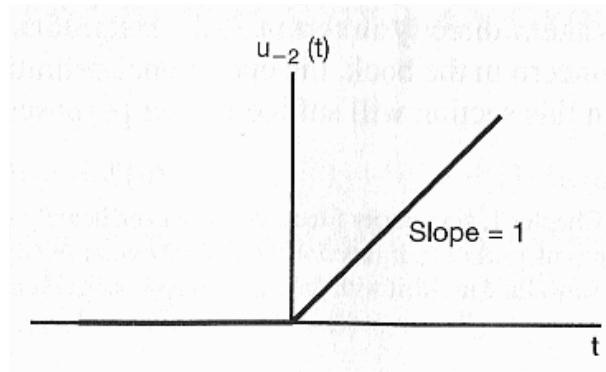
- Unit Doublets of Integration Operation:

- Similarly,

$$u_{-2}(t) = u(t) * u(t) = \int_{-\infty}^t u(\tau) d\tau$$

- That is,

$$u_{-2}(t) = t u(t) \quad \text{the unit ramp function}$$



- Unit Doublets of Integration Operation:

- Moreover,

$$x(t) * u_{-2}(t) = x(t) * u(t) * u(t)$$

$$= \left(\int_{-\infty}^t x(\sigma) d\sigma \right) * u(t)$$

$$= \int_{-\infty}^t \left(\int_{-\infty}^{\tau} x(\sigma) d\sigma \right) d\tau$$

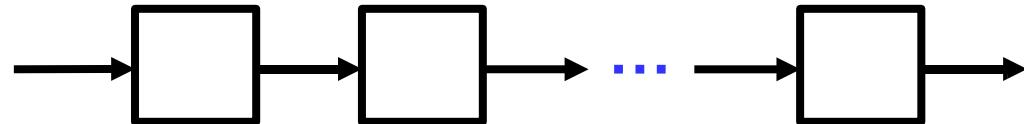
- In general,

$$u_{-k}(t) = u(t) * \cdots * u(t) = \int_{-\infty}^t u_{-(k-1)}(\tau) d\tau$$

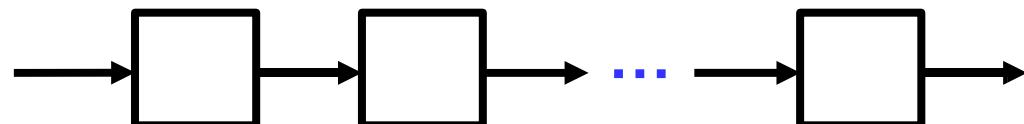
$$u_{-k}(t) = \frac{t^{k-1}}{(k-1)!} u(t)$$

■ In Summary

$$\delta(t) = u_0(t)$$



$$u(t) = u_{-1}(t)$$



$$u_k(t)$$

$$k > 0,$$

Impulse response of a cascade of k differentiators

$$k < 0,$$

Impulse response of a cascade of $|k|$ integrators

$$u(t) * u_1(t) = \delta(t) \quad \text{or, } u_{-1}(t) * u_1(t) = u_0(t)$$

$$\Rightarrow u_k(t) * u_r(t) = u_{k+r}(t)$$