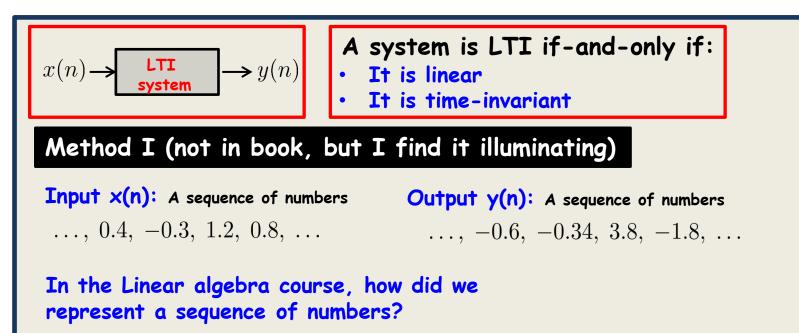
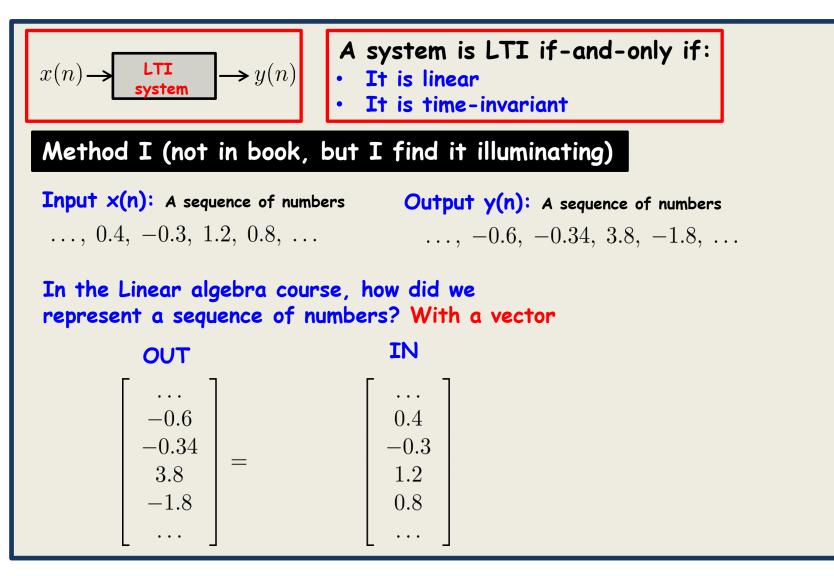
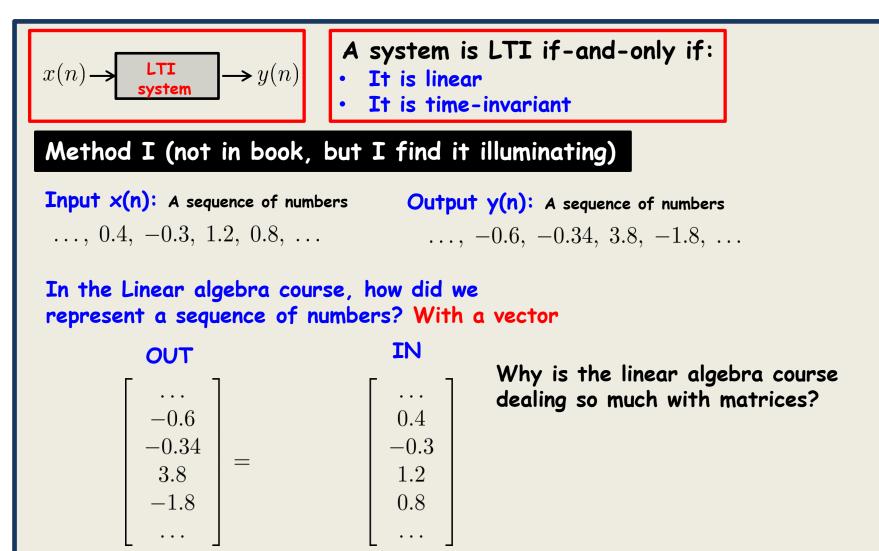
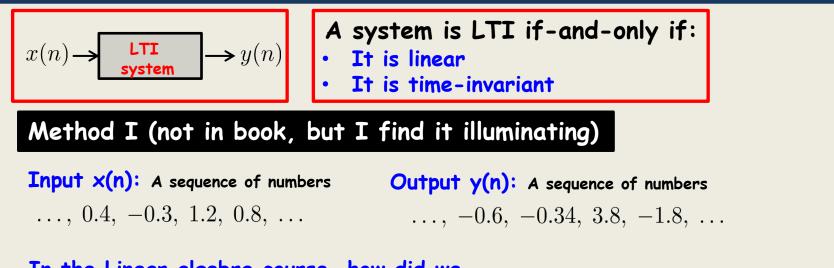


LTI systems have compact mathematical representation We next provide two ways the reach the representation

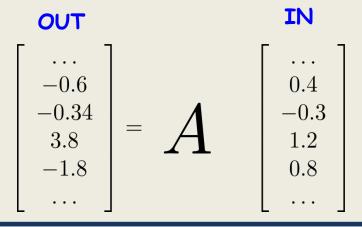




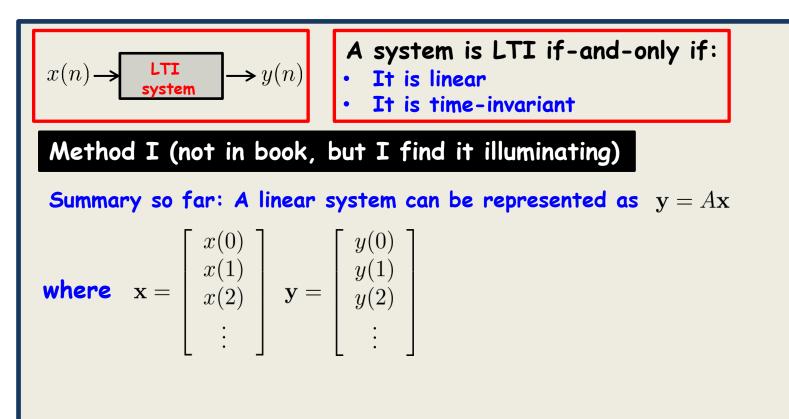


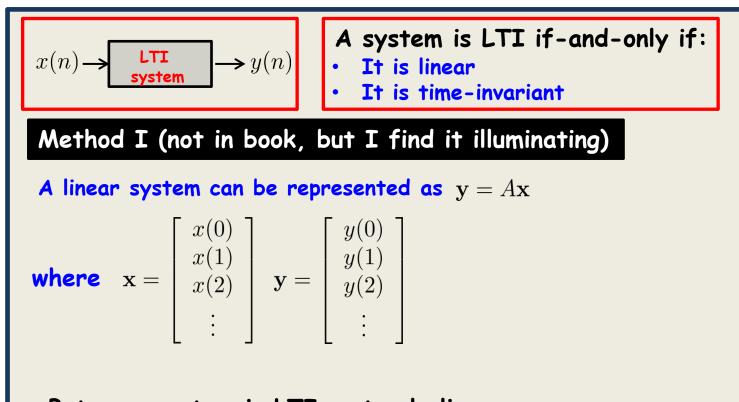


In the Linear algebra course, how did we represent a sequence of numbers? With a vector

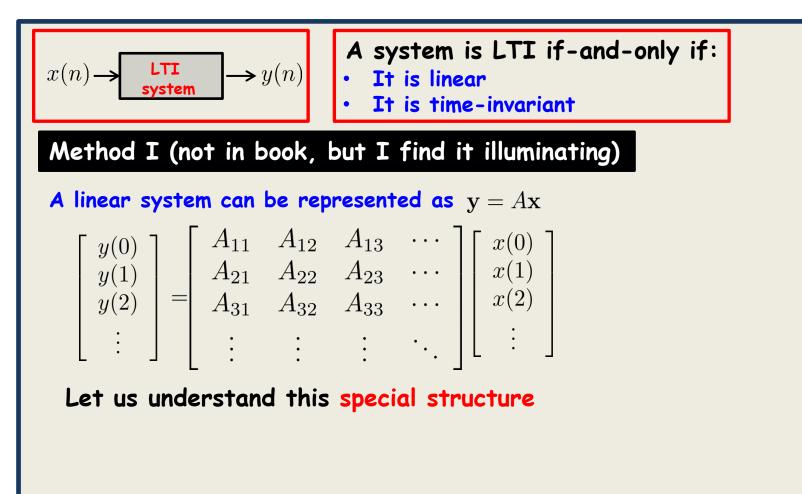


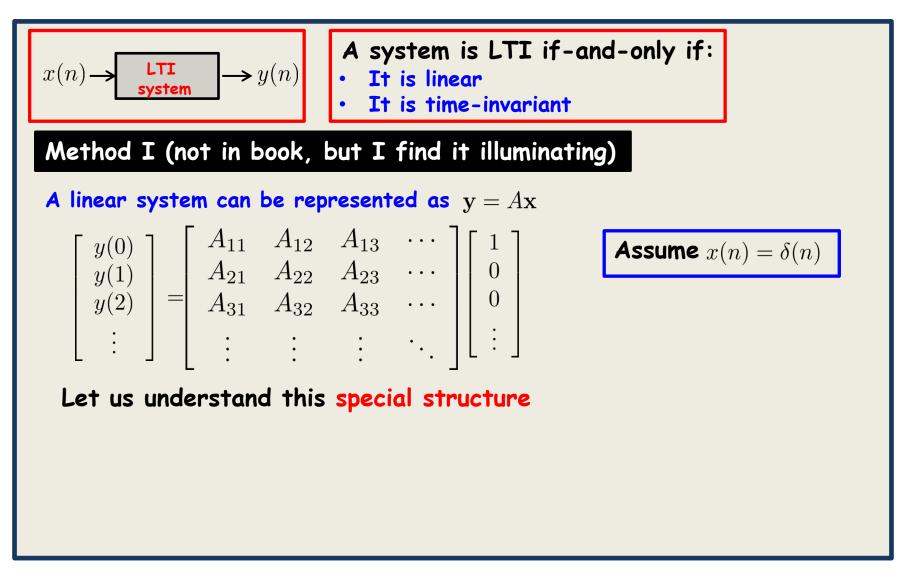
Why is the linear algebra course dealing so much with matrices? Because every linear function can be represented by a matrix

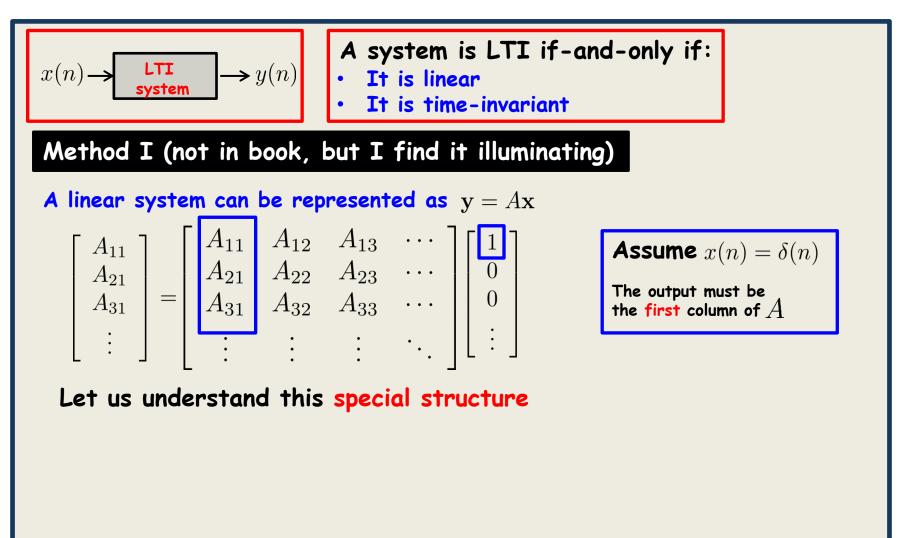


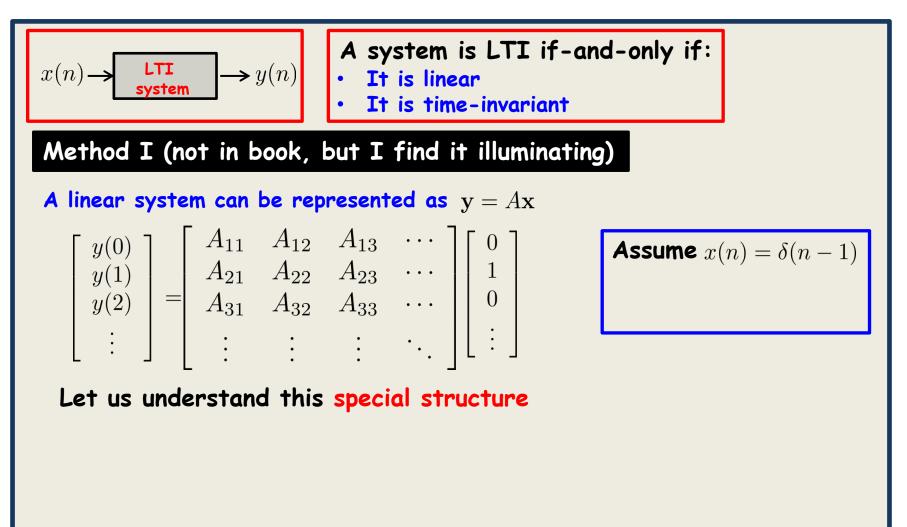


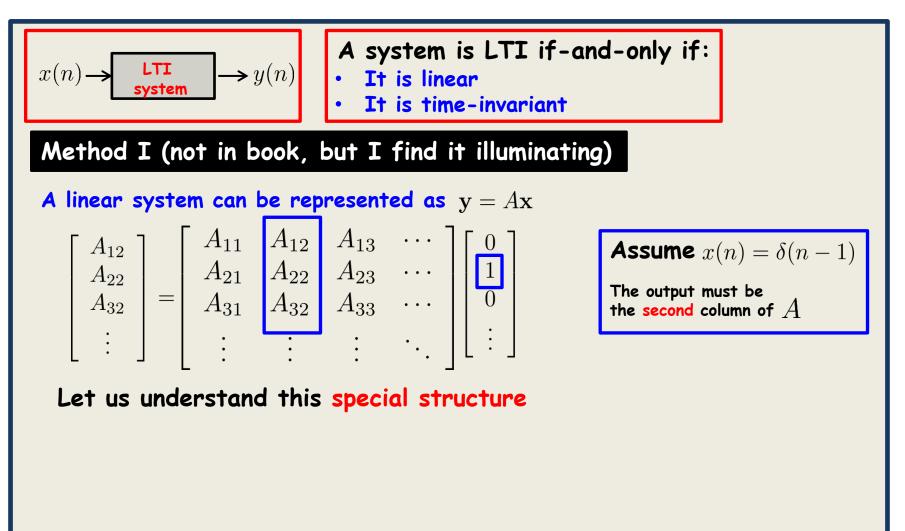
But, our system is LTI, not only linear, so this imposes restrictions on A i.e., A must have a special structure

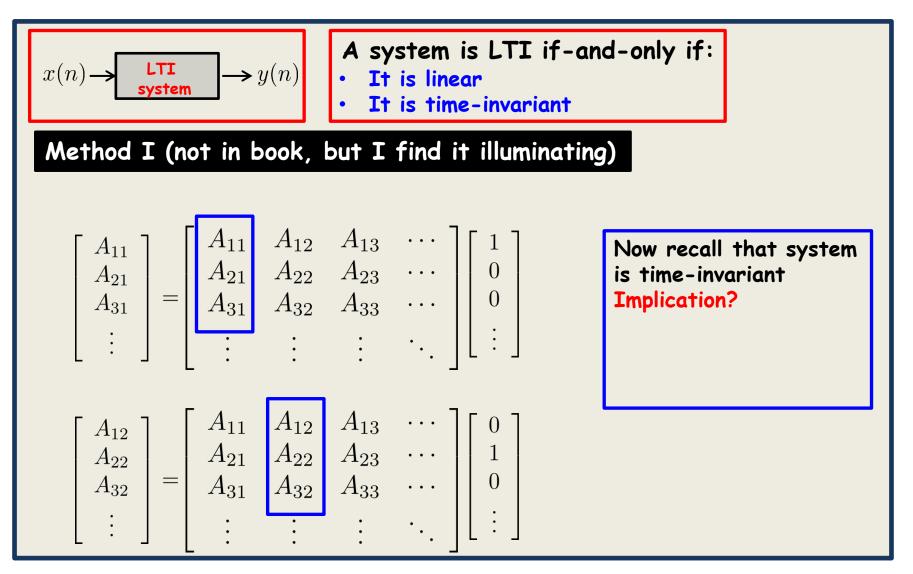


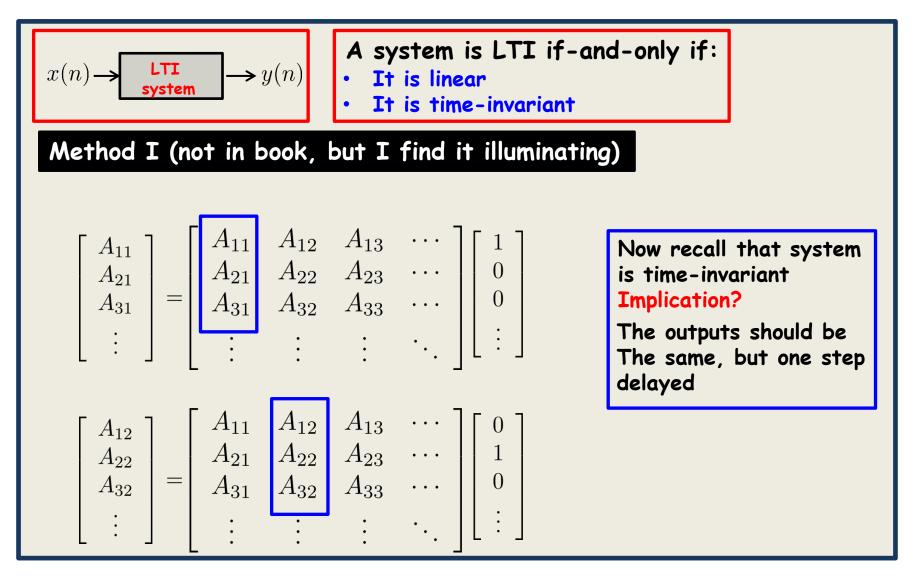


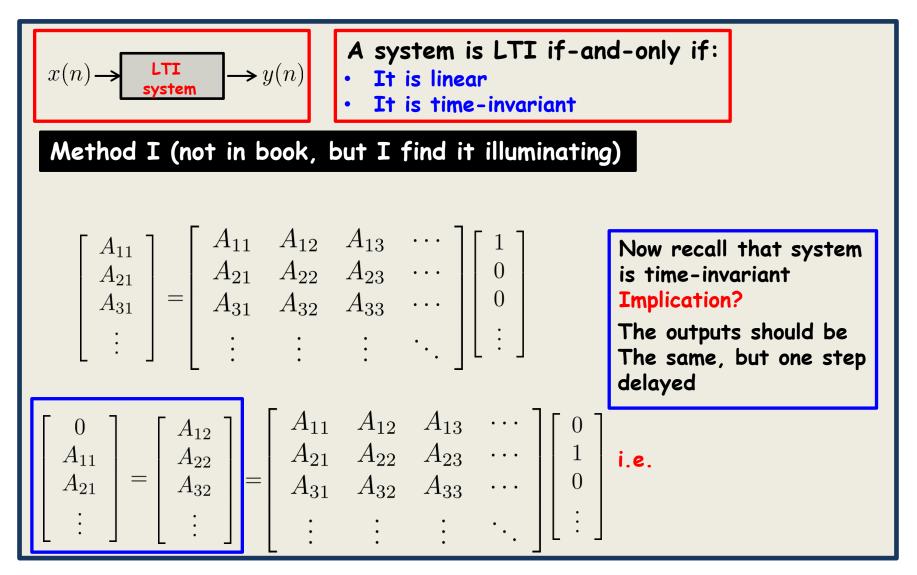


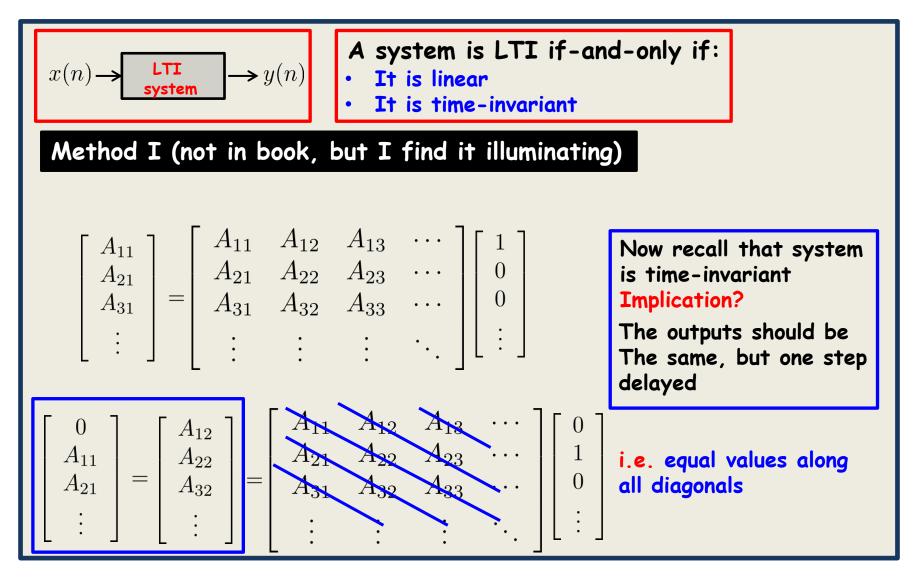


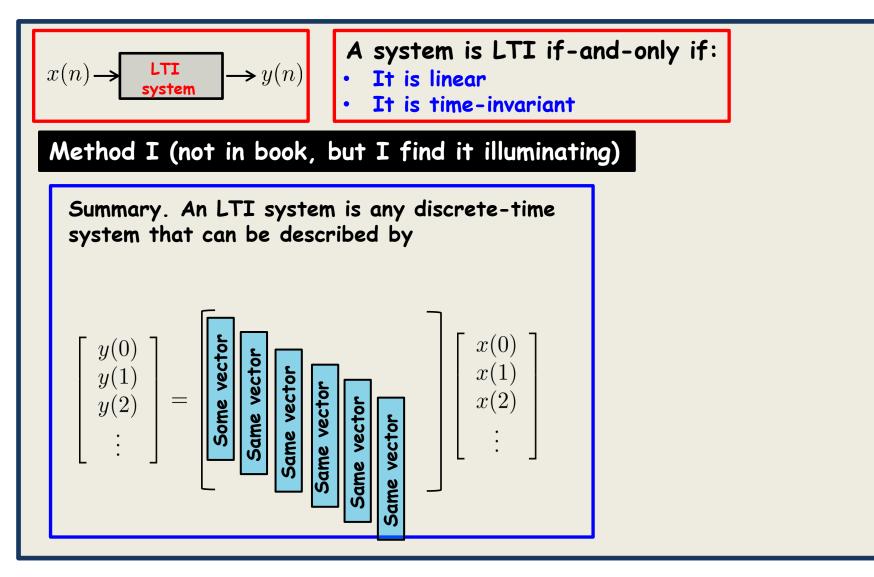


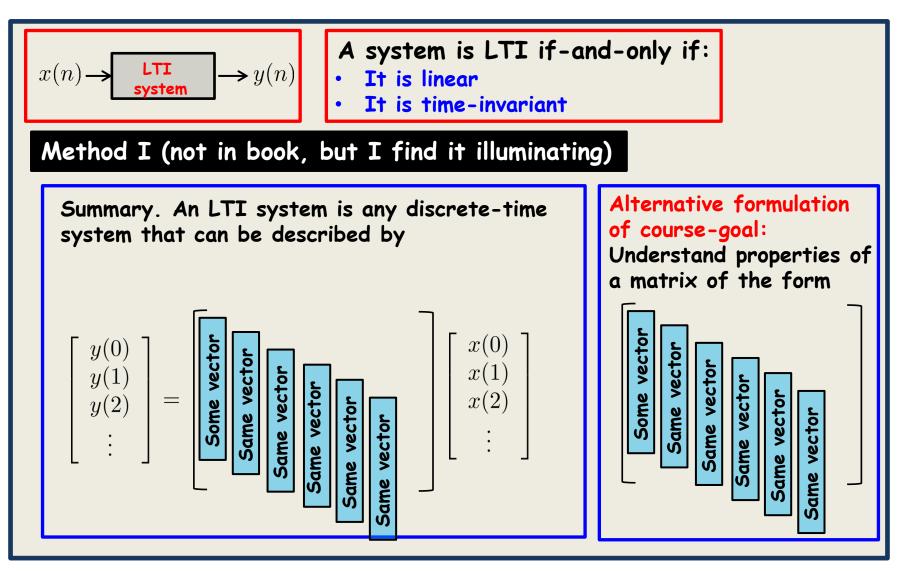


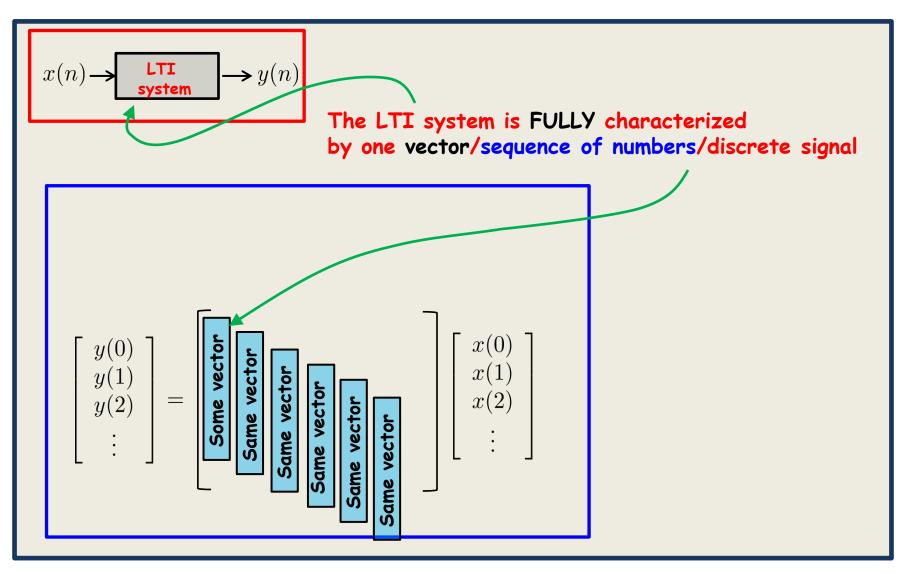


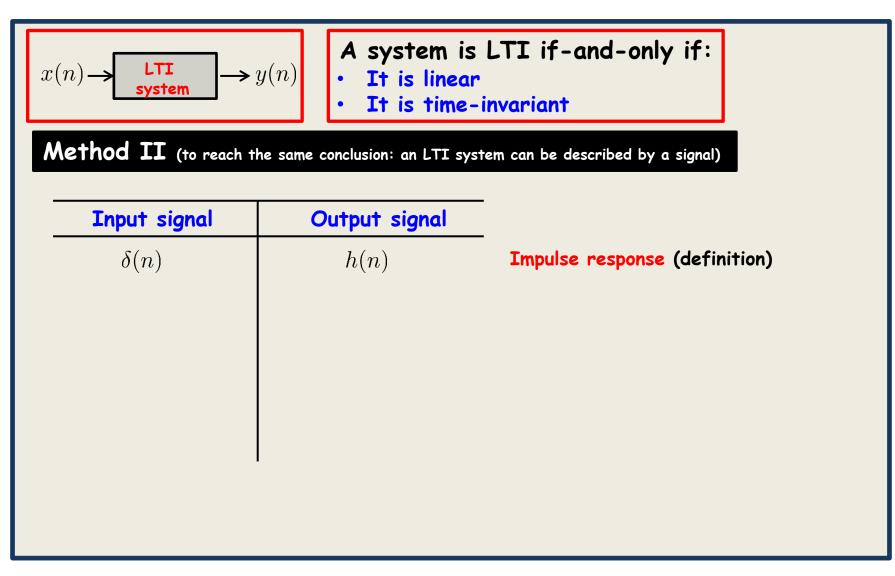


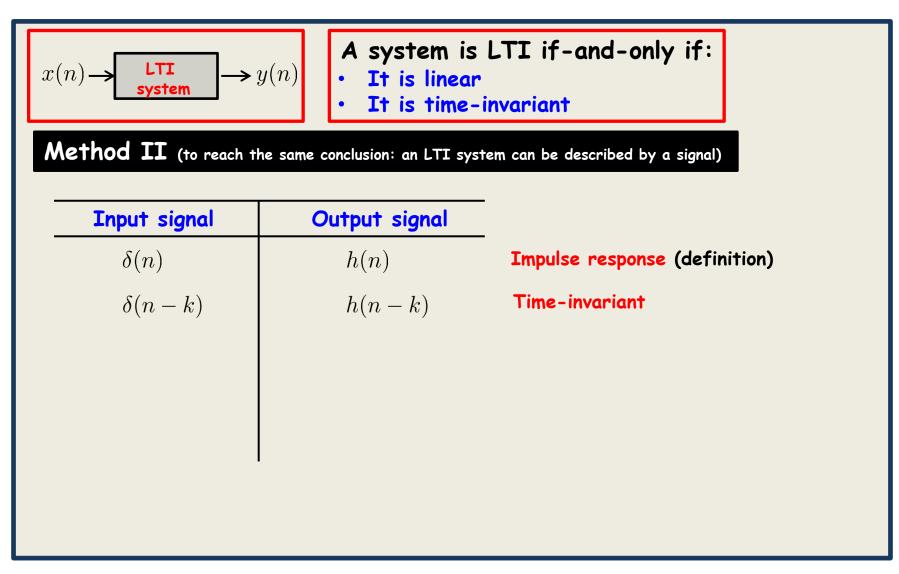


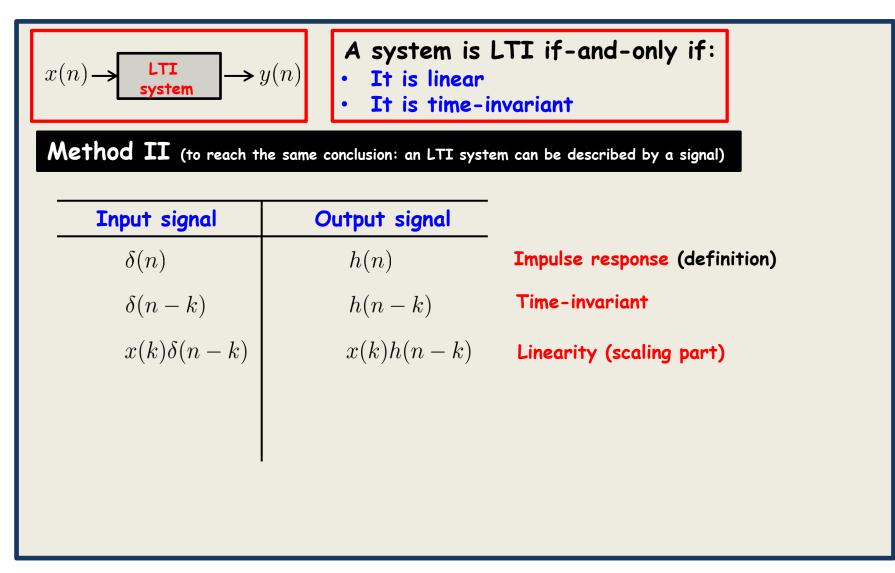


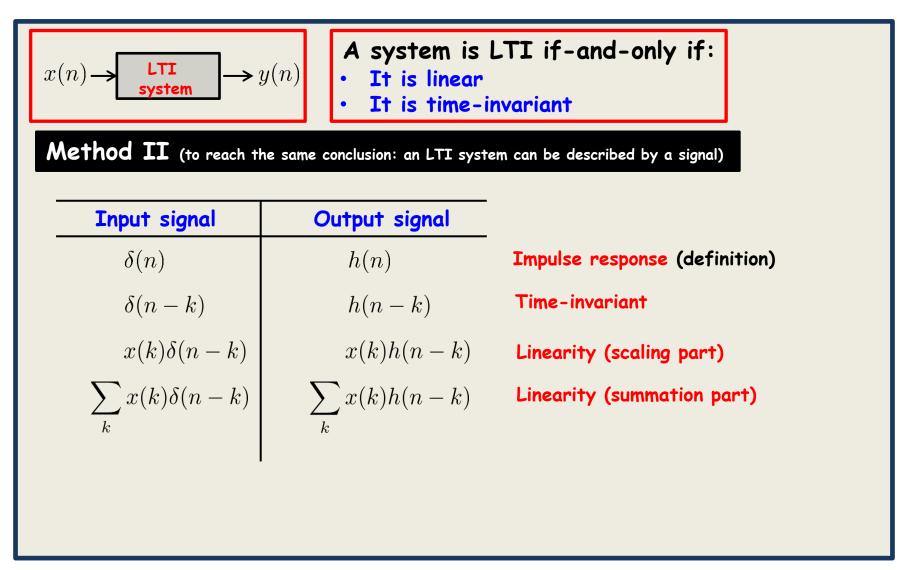


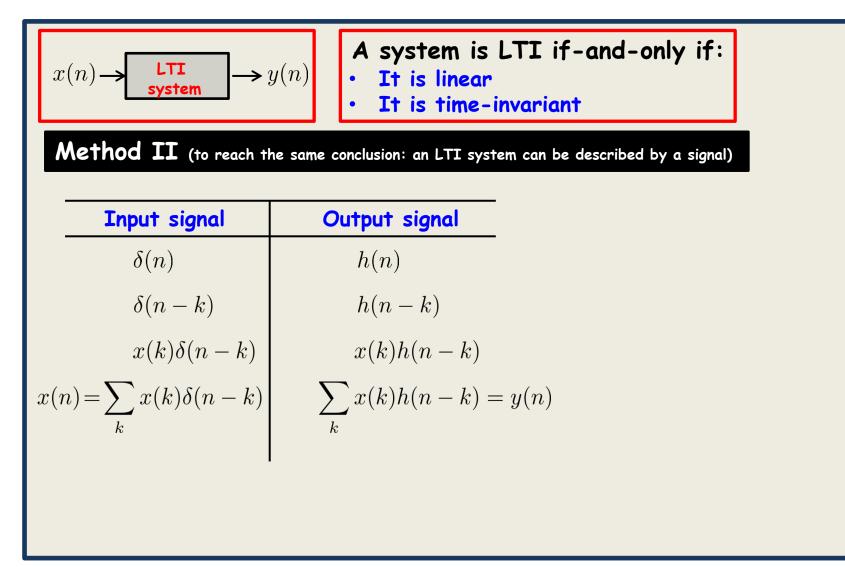


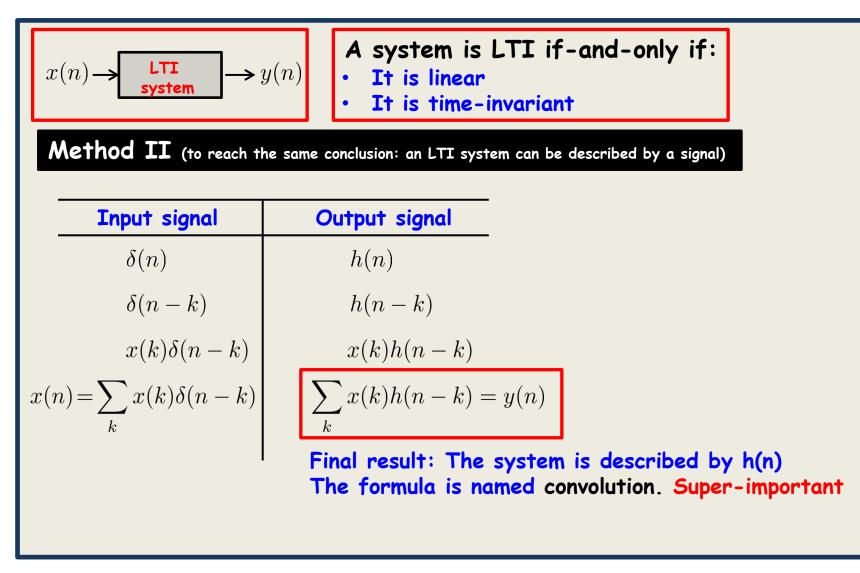


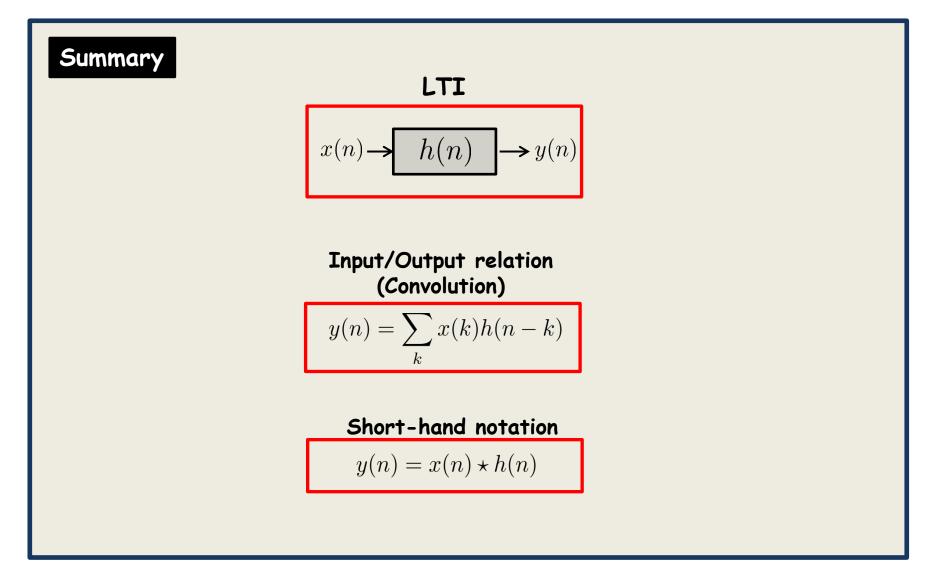












#### Agenda

#### Today

Get familiar with  $y(n) = x(n) \star h(n)$  through some examples For what h(n) do we have BIBO stability? See relationship between h(n) and  $\sum_{k} a(k)y(n-k) = \sum_{\ell} b(\ell)x(n-\ell)$ Some notes on correlation functions

#### In the long run

#### Agenda

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In the long run (Loosely speaking)

 $\begin{array}{l} {\rm Study}\sum_k a(k)y(n-k) = \sum_\ell b(\ell)x(n-\ell) \ \ {\rm in \ \ detail \ \ via \ \ z-transform, \ \ and \ \ 2 \ \ types} \\ {\rm of \ \ Fourier \ \ transforms} \end{array}$ 

The sampling-reconstruction issues

#### Agenda

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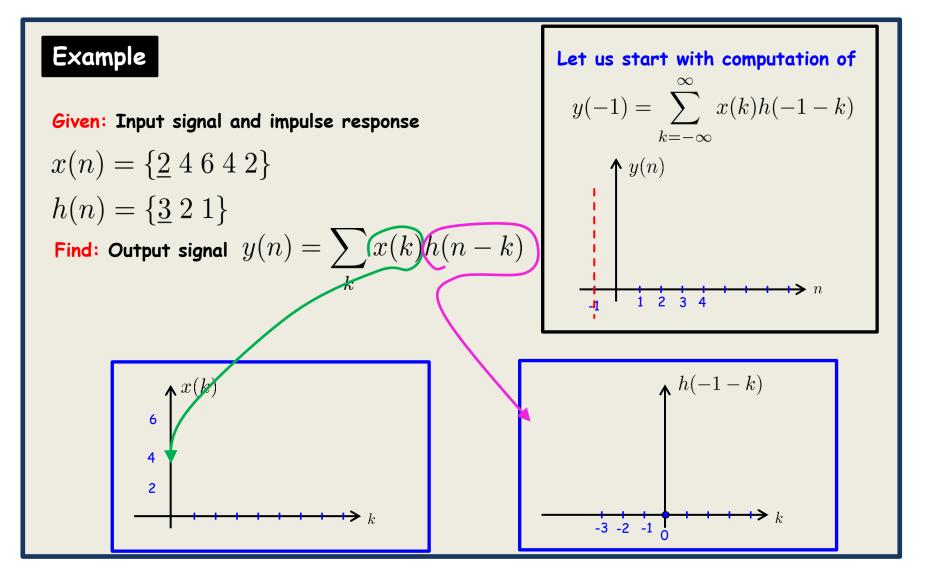
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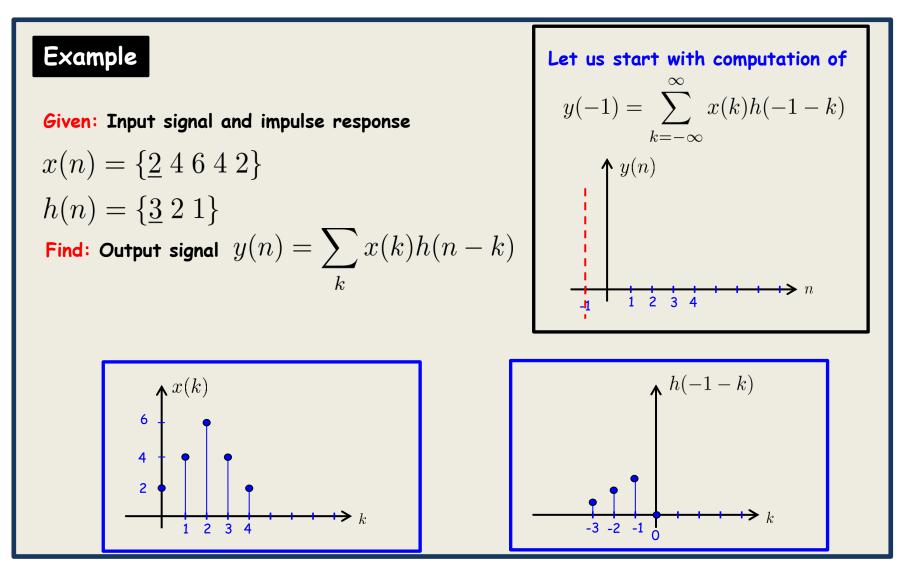
The sampling-reconstruction issues

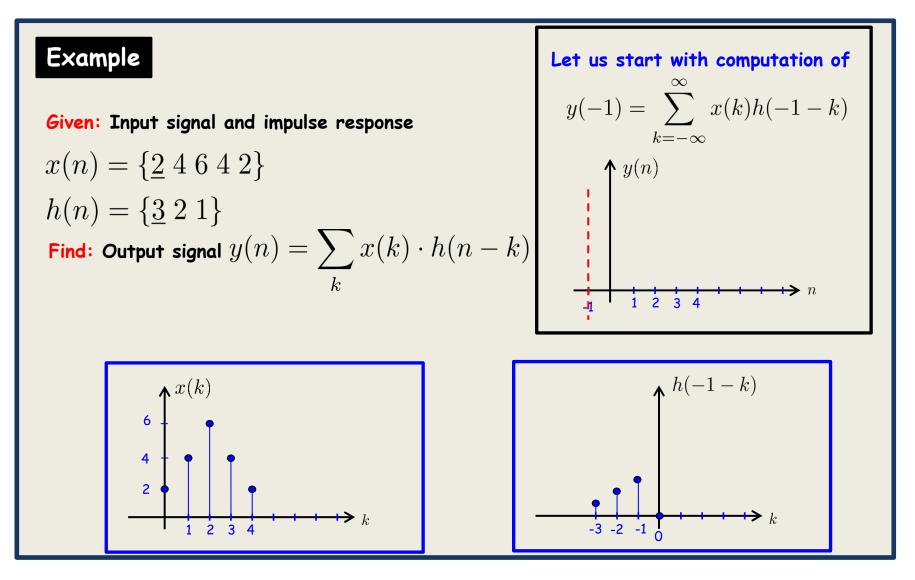
#### Example

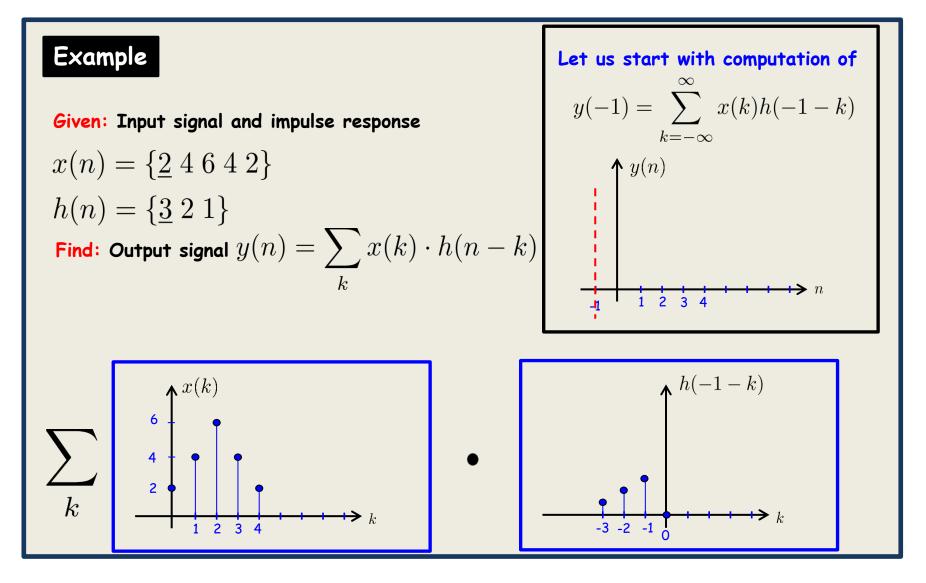
Given: Input signal and impulse response  $x(n) = \{ \underline{2} \ 4 \ 6 \ 4 \ 2 \}$   $h(n) = \{ \underline{3} \ 2 \ 1 \}$ Find: Output signal  $y(n) = \sum_{k} x(k)h(n-k)$ 

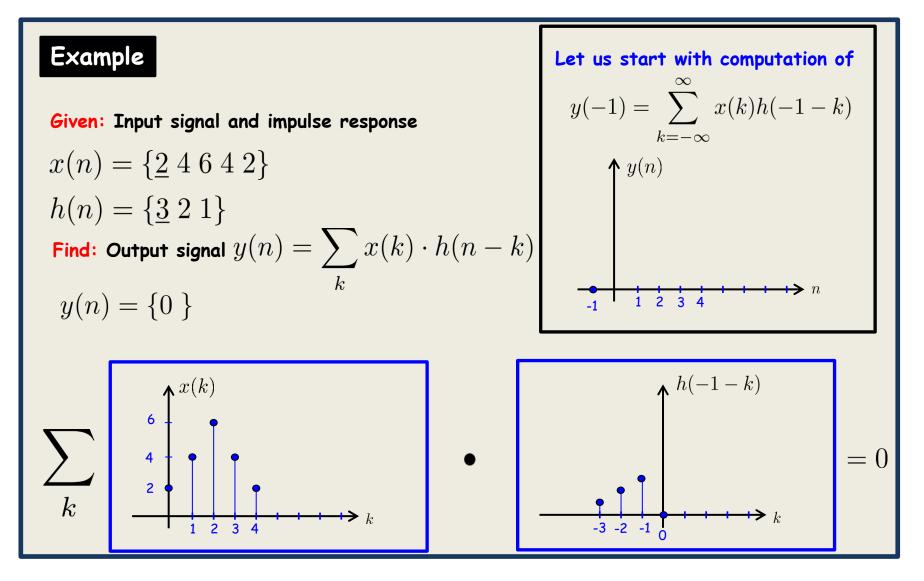
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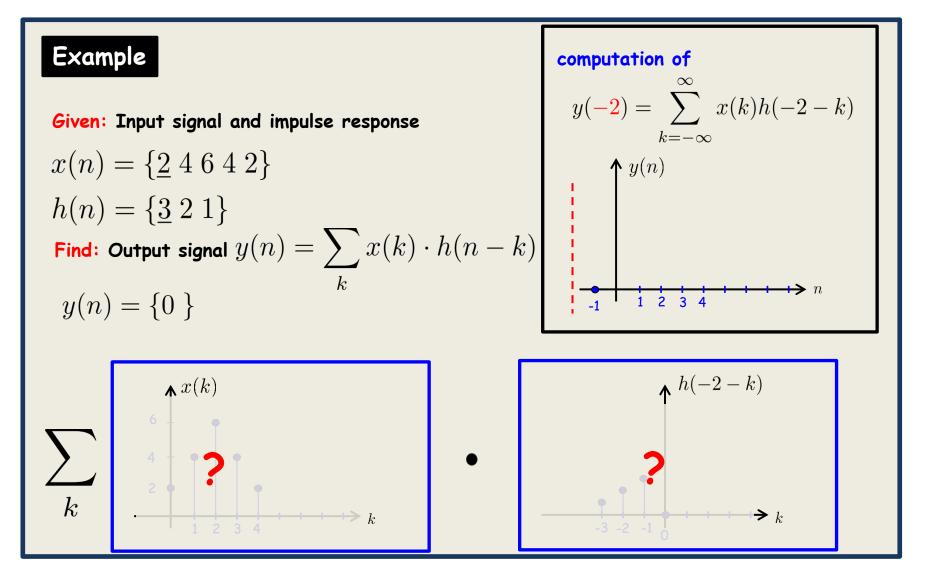


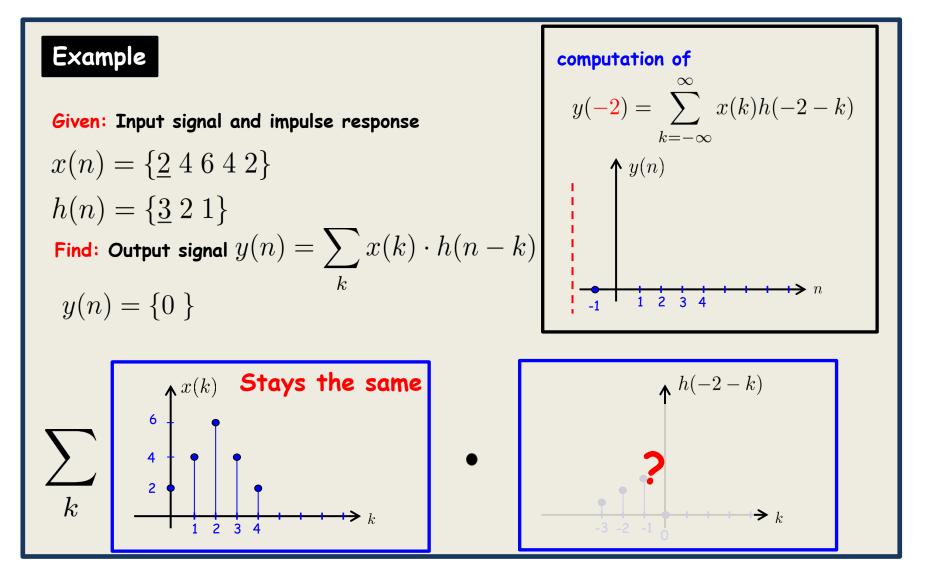


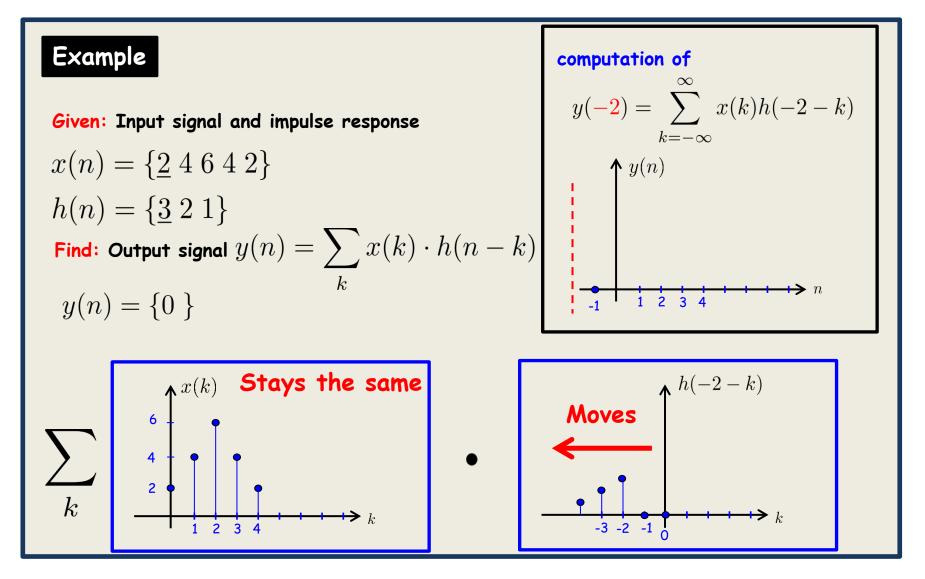


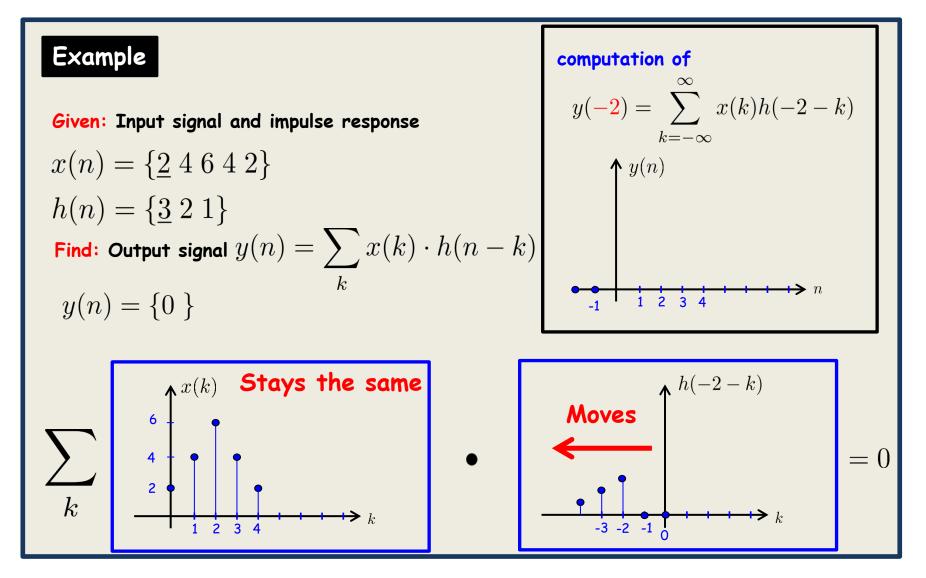


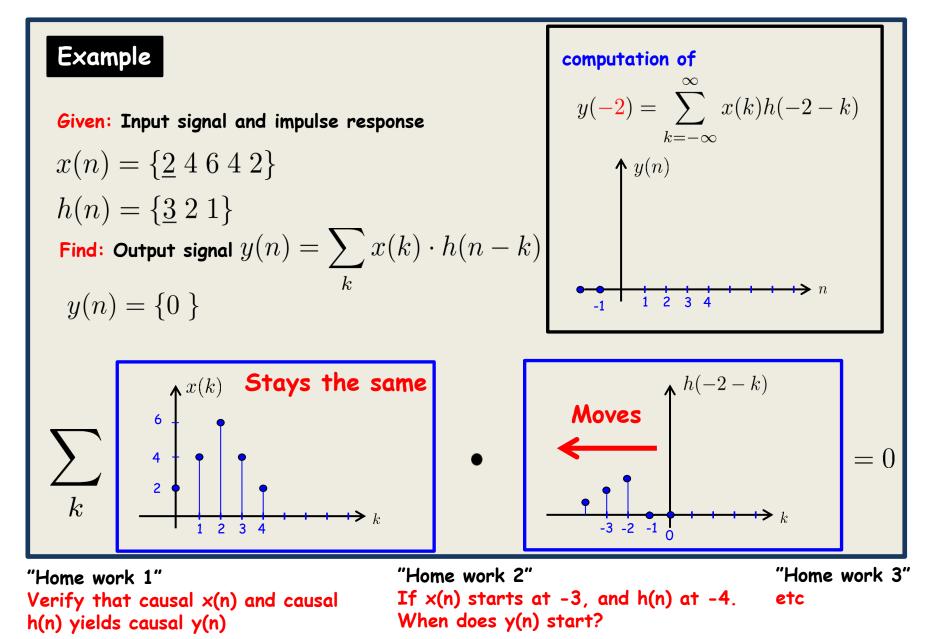


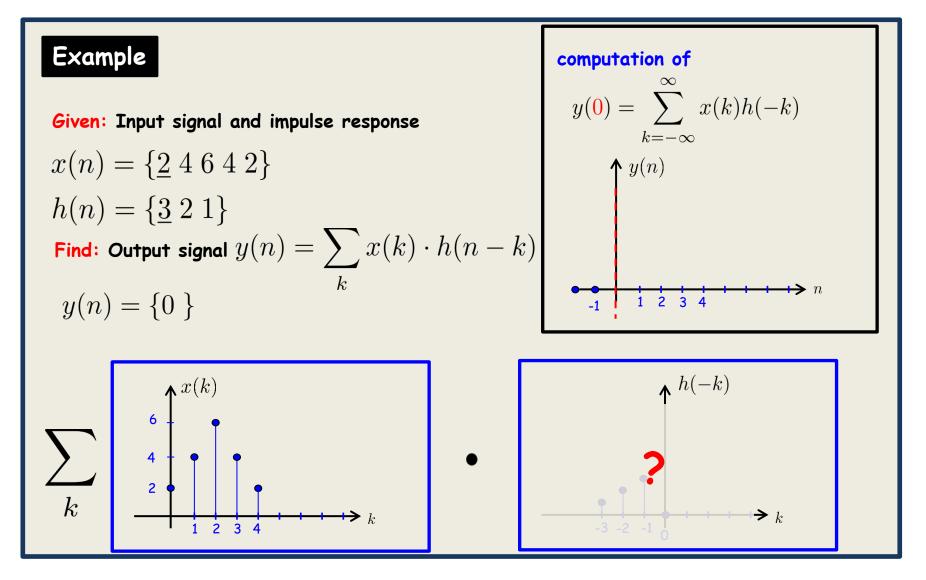


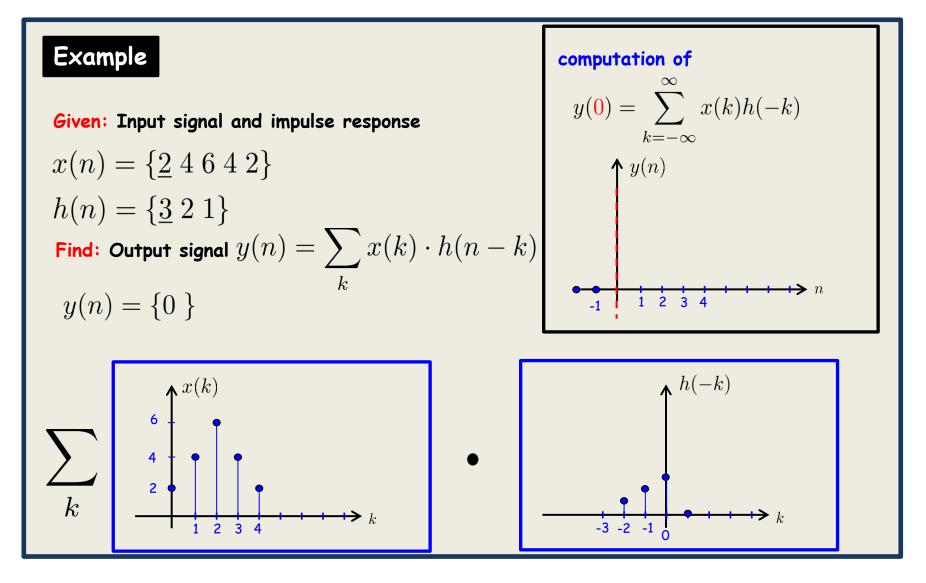


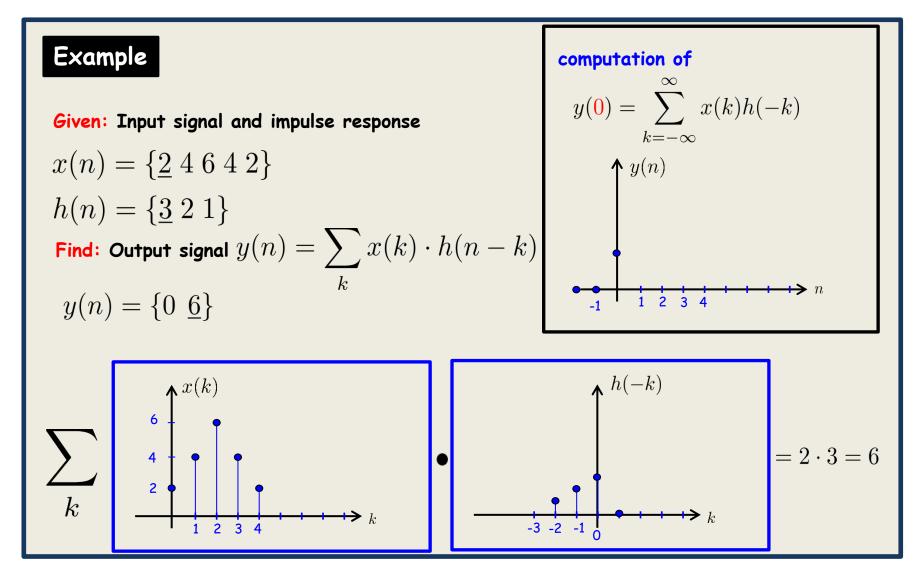


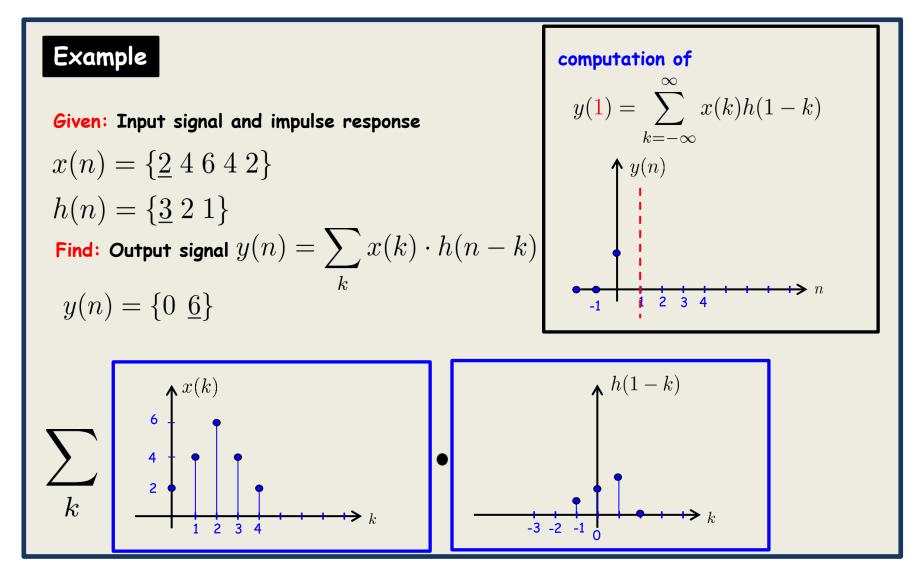


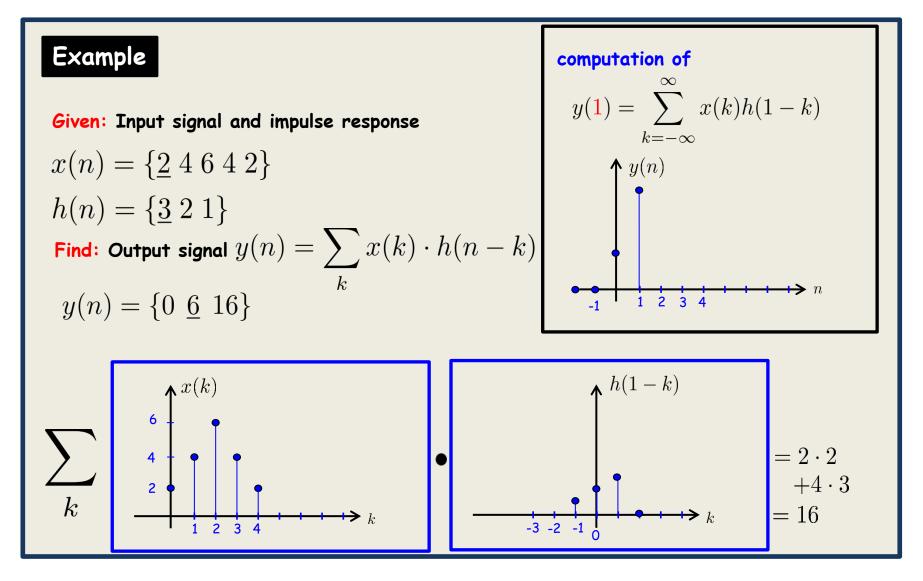


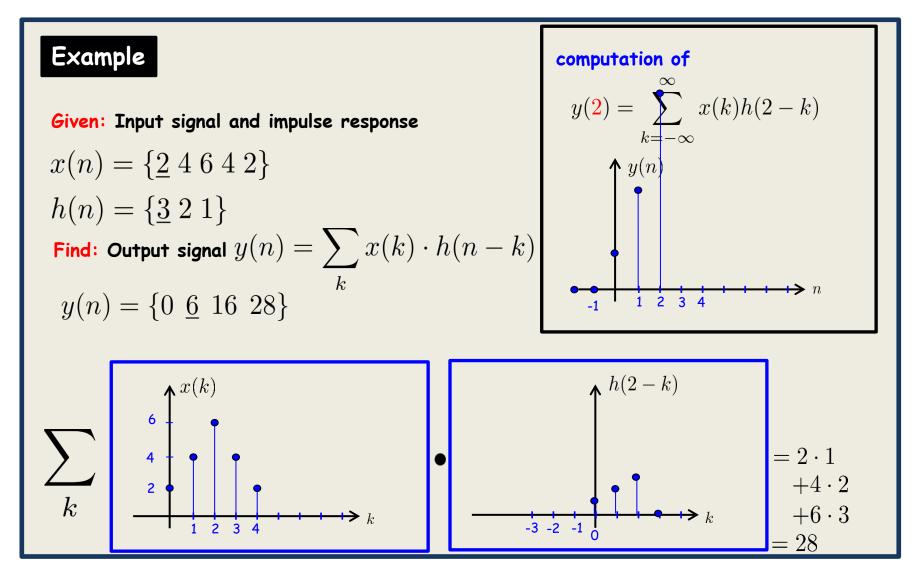












#### Example

Given: Input signal and impulse response  $x(n) = \{\underline{2} \ 4 \ 6 \ 4 \ 2\}$   $h(n) = \{\underline{3} \ 2 \ 1\}$ Find: Output signal  $y(n) = \sum_{k} x(k) \cdot h(n-k)$  $y(n) = \{0 \ \underline{6} \ 16 \ 28 \ 28 \ \underline{20} \ 8 \ 2\}$ 

**Repeating gives** 

#### Example

Given: Input signal and impulse response  $x(n) = \{\underline{2} \ 4 \ 6 \ 4 \ 2\}$   $h(n) = \{\underline{3} \ 2 \ 1\}$ Find: Output signal  $y(n) = \sum_{k} x(k) \cdot h(n-k)$  $y(n) = \{\underline{6} \ 16 \ 28 \ 28 \ 20 \ 8 \ 2\}$ 

#### Example

#### Three more methods

Given: Input signal and impulse response  $x(n) = \{\underline{2} \ 4 \ 6 \ 4 \ 2\}$   $h(n) = \{\underline{3} \ 2 \ 1\}$ Find: Output signal  $y(n) = \sum_{k} x(k) \cdot h(n-k)$ 

#### Example

#### Three more methods Method 1

Given: Input signal and impulse response  $x(n) = \{\underline{2} \ 4 \ 6 \ 4 \ 2\}$   $h(n) = \{\underline{3} \ 2 \ 1\}$ Find: Output signal  $y(n) = \sum_{k} x(k) \cdot h(n-k)$  = 3x(n) + 2x(n-1) + x(n-2)

#### Example

#### Three more methods Method 1

Given: Input signal and impulse response  $x(n) = \{\underline{2} \ 4 \ 6 \ 4 \ 2\}$  $h(n) = \{\underline{3} \ 2 \ 1\}$ Find: Output signal  $y(n) = \sum x(k) \cdot h(n-k)$ = 3x(n) + 2x(n-1) + x(n-2)n = 0 $h(0-k) \qquad 1 \quad 2 \quad \underline{3}$ 2 4 6 4 2 x(k)h(0-k)x(k) $\sum = 6 = \psi(0)$ 6

#### Example

#### Three more methods Method 1

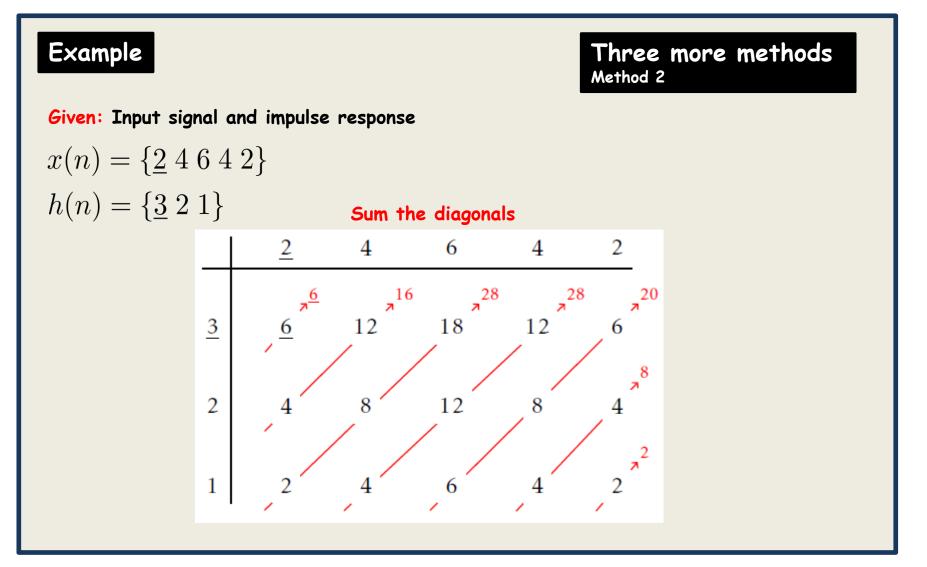
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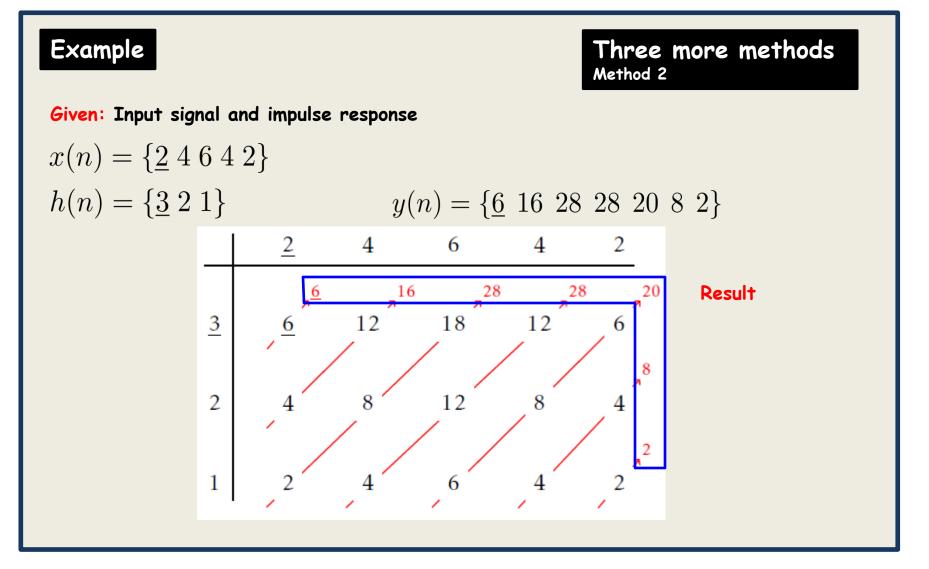
#### Example

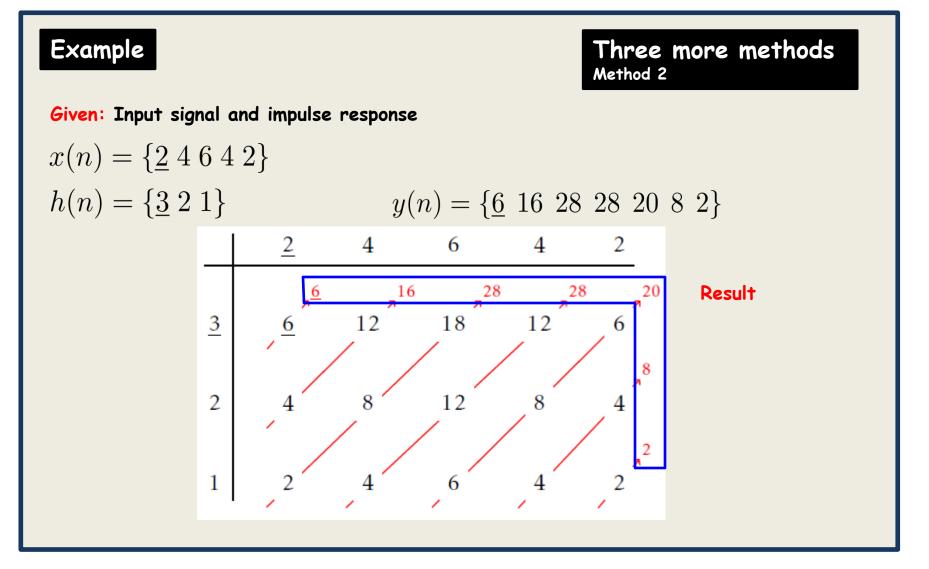
#### Three more methods Method 1

Given: Input signal and impulse response  $x(n) = \{\underline{2} \ 4 \ 6 \ 4 \ 2\}$  $h(n) = \{\underline{3} \ 2 \ 1\}$ Find: Output signal  $y(n) = \sum x(k) \cdot h(n-k)$ = 3x(n) + 2x(n-1) + x(n-2)n=2h(2-k)1 2 <u>3</u> 2 4 6 4 2 x(k) $\sum = 28 = y(2)$ h(2-k)x(k)2 8 18

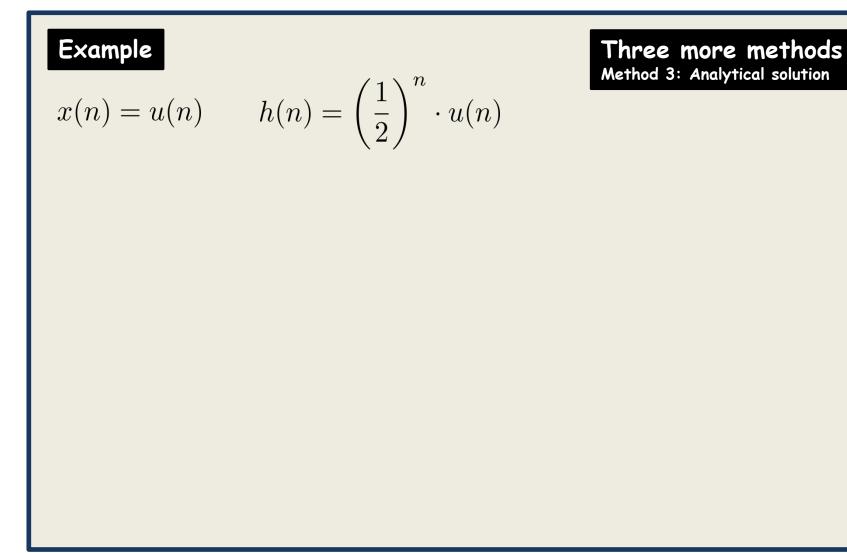
#### Example Three more methods Method 2 Given: Input signal and impulse response $x(n) = \{\underline{2} \ 4 \ 6 \ 4 \ 2\}$ $h(n) = \{\underline{3} \ 2 \ 1\} \qquad \text{Put numbers in a table and multiply}$ 2 4 6 4 2 <u>6</u> 12 18 12 6 3 4 8 12 8 4 2 2 4 6 4 2 1

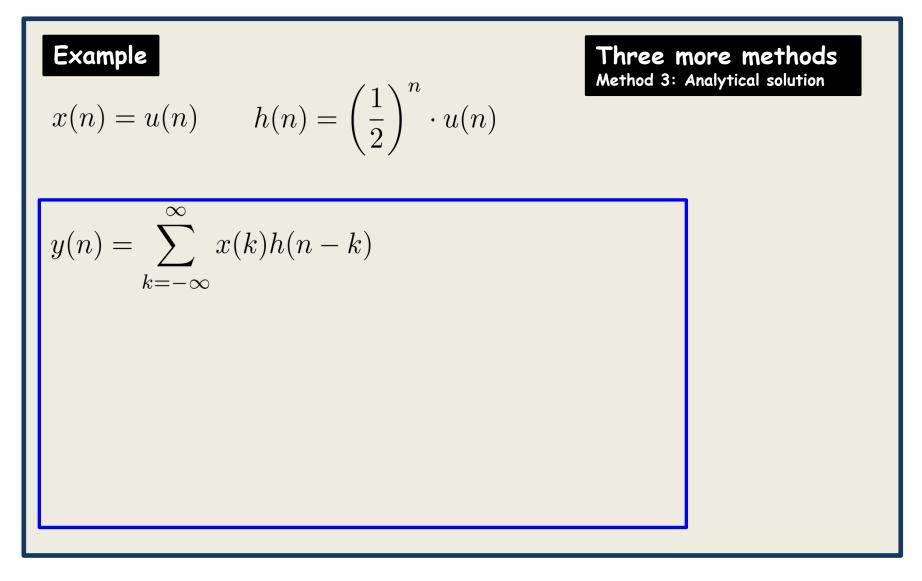






Make sure that you understand why a convolution of a length K signal with a length L signal has length K+L-1





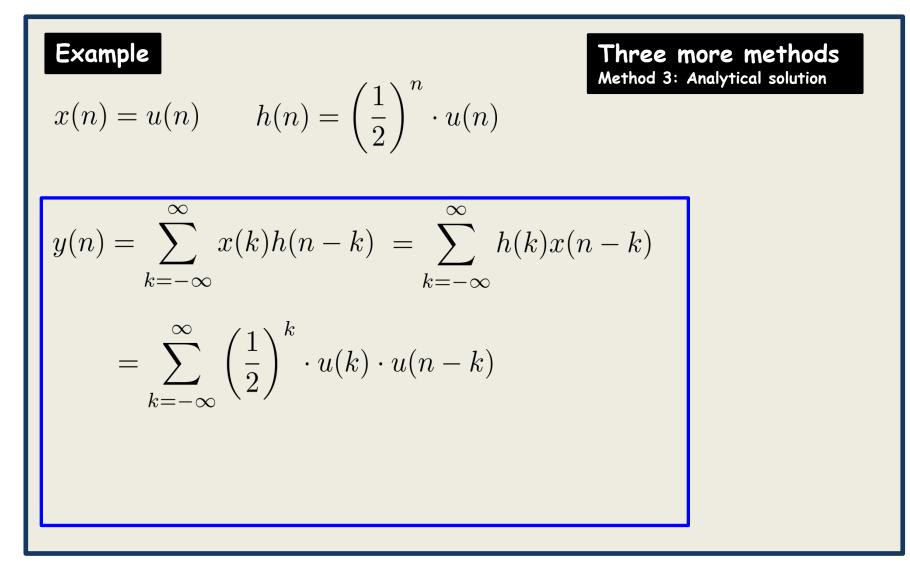
#### Example

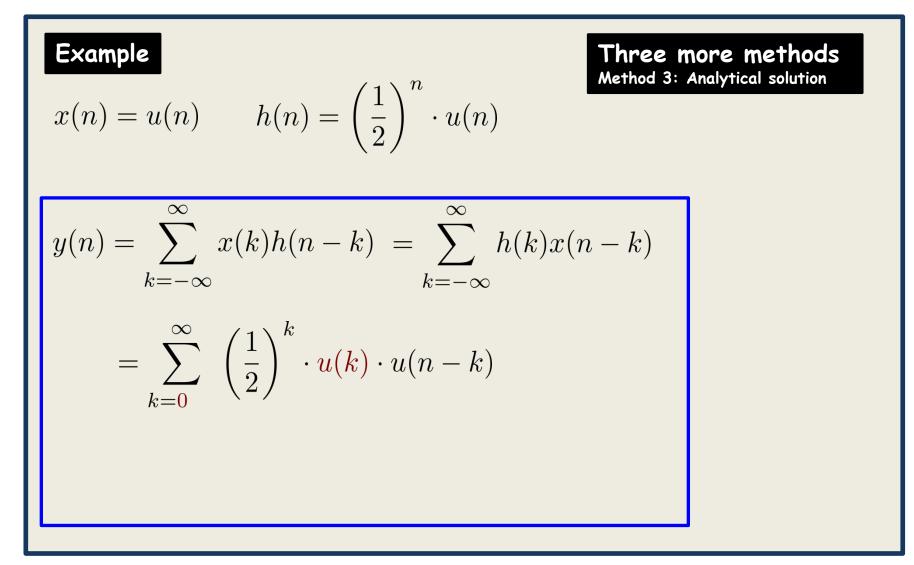
$$x(n) = u(n)$$
  $h(n) = \left(\frac{1}{2}\right)^n \cdot u(n)$ 

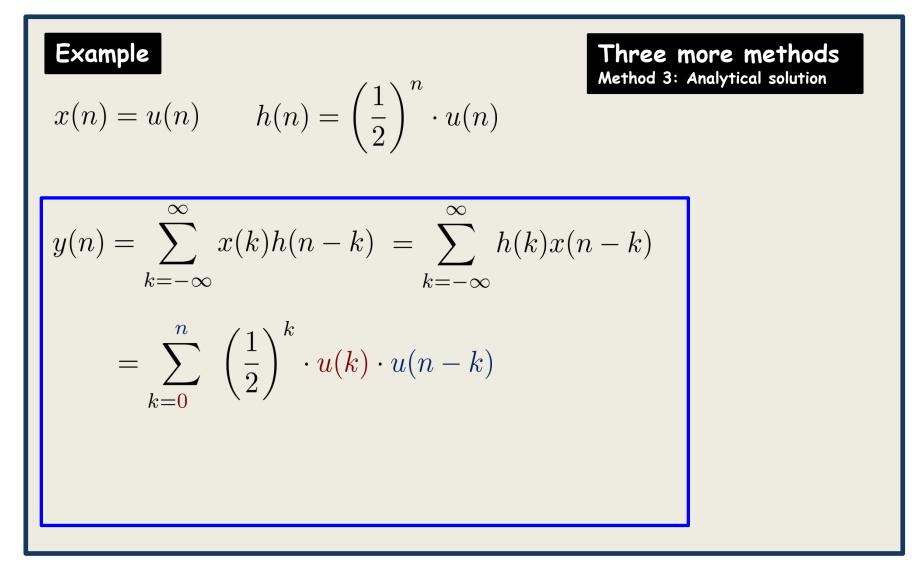
Three more methods Method 3: Analytical solution

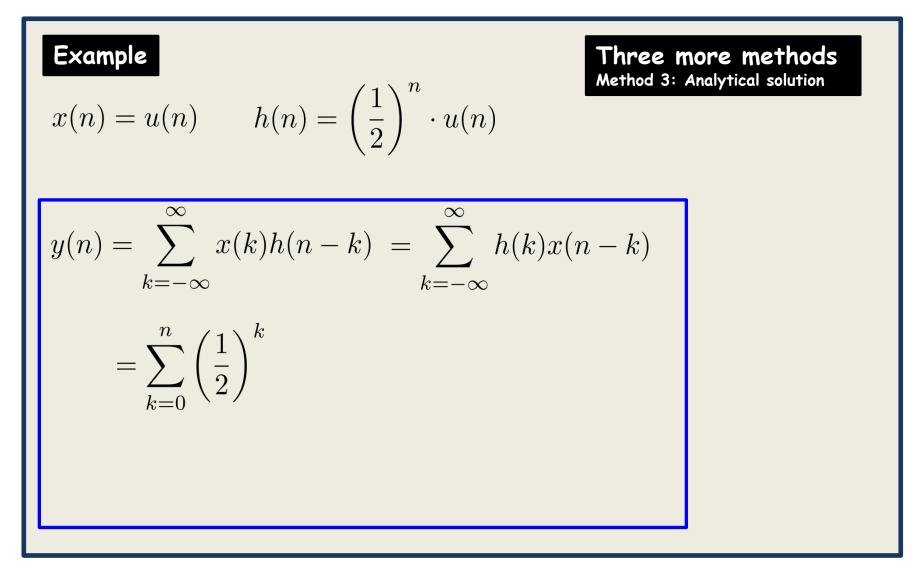
$$y(n) = \sum_{k=-\infty}^{\infty} \widehat{x(k)h(n-k)} = \sum_{k=-\infty}^{\infty} h(k)x(n-k)$$

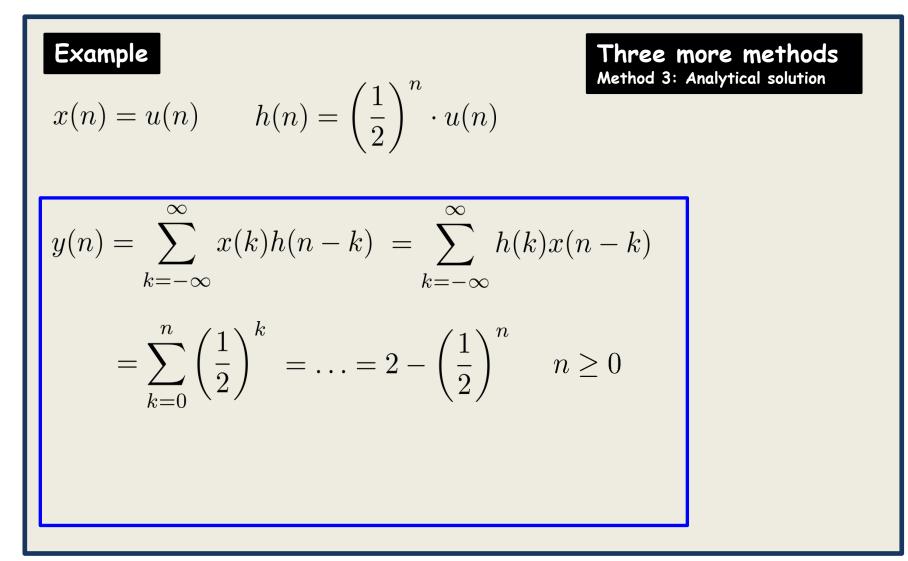
Minor trick. Not really needed, but slightly simpler. Try without the trick at home

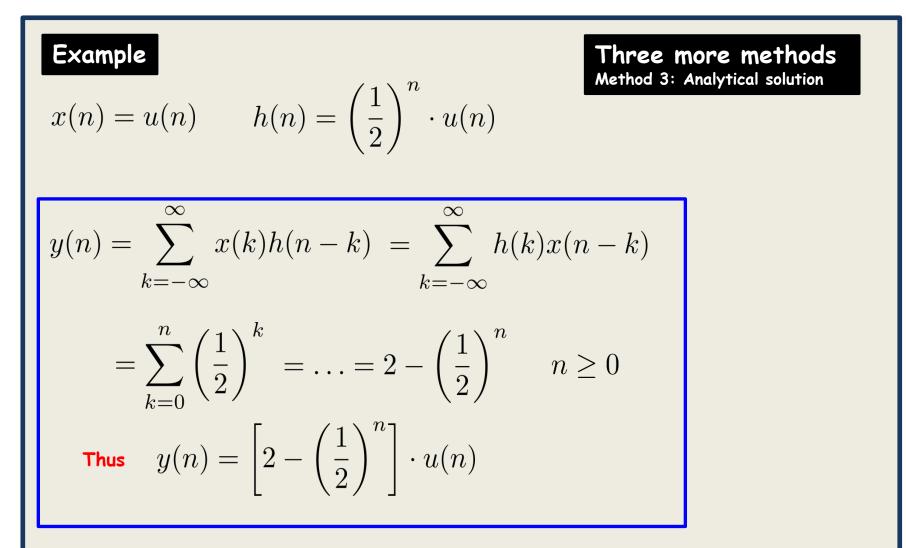












#### Standard Properties

#### Commutativity

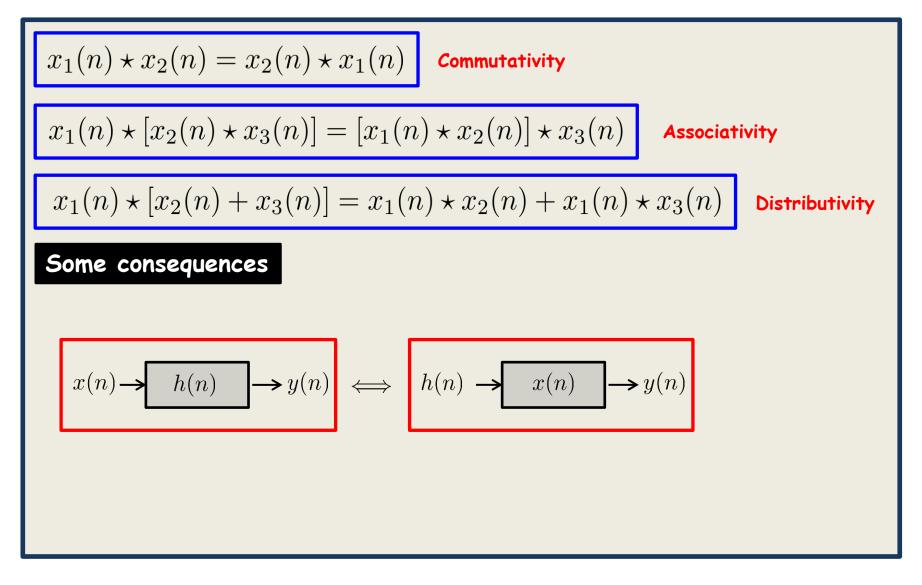
$$x_1(n) \star x_2(n) = x_2(n) \star x_1(n)$$

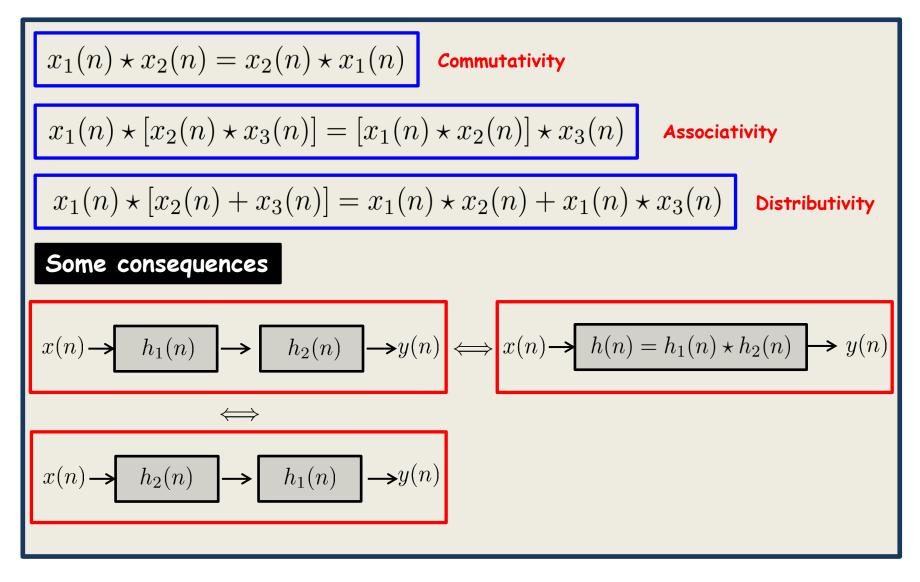
#### Associativity

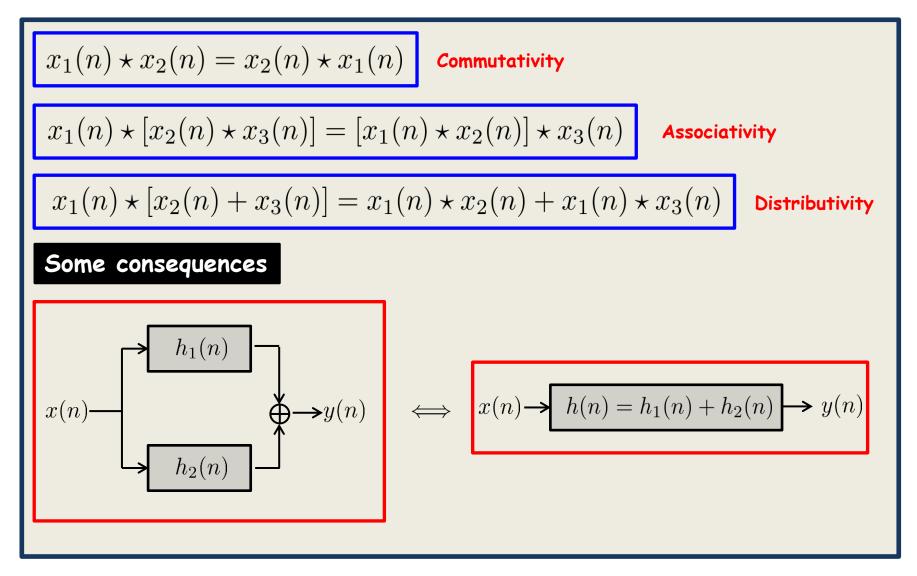
$$x_1(n) \star [x_2(n) \star x_3(n)] = [x_1(n) \star x_2(n)] \star x_3(n)$$

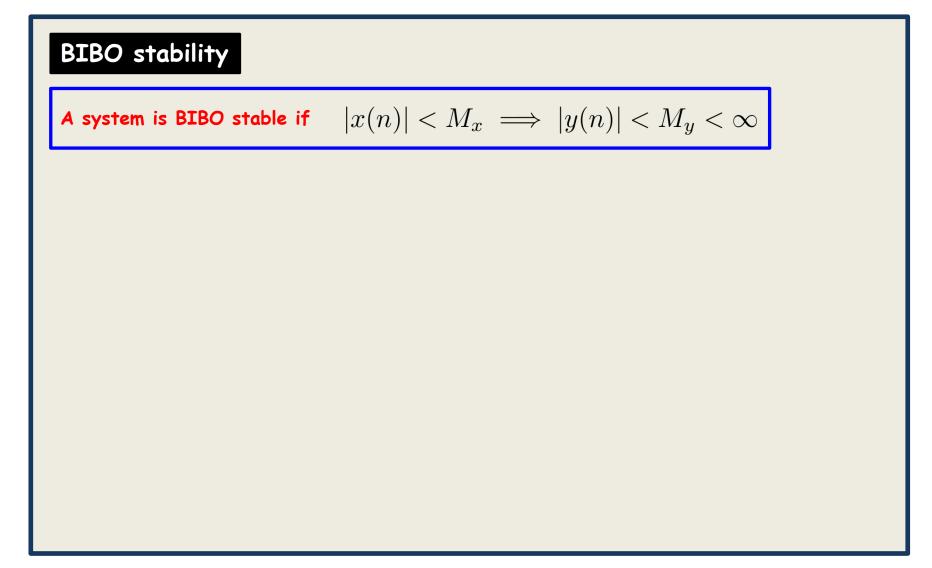
#### Distributivity

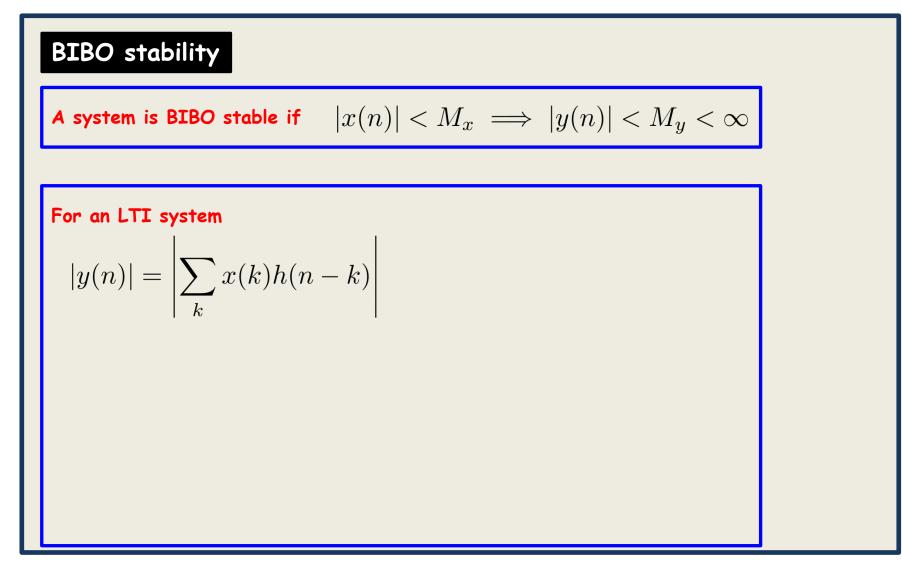
$$x_1(n) \star [x_2(n) + x_3(n)] = x_1(n) \star x_2(n) + x_1(n) \star x_3(n)$$

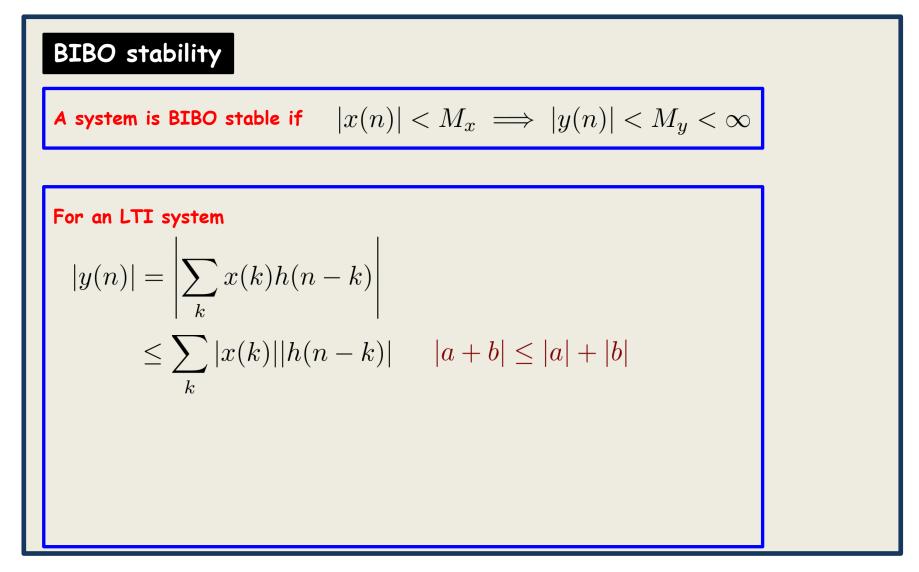


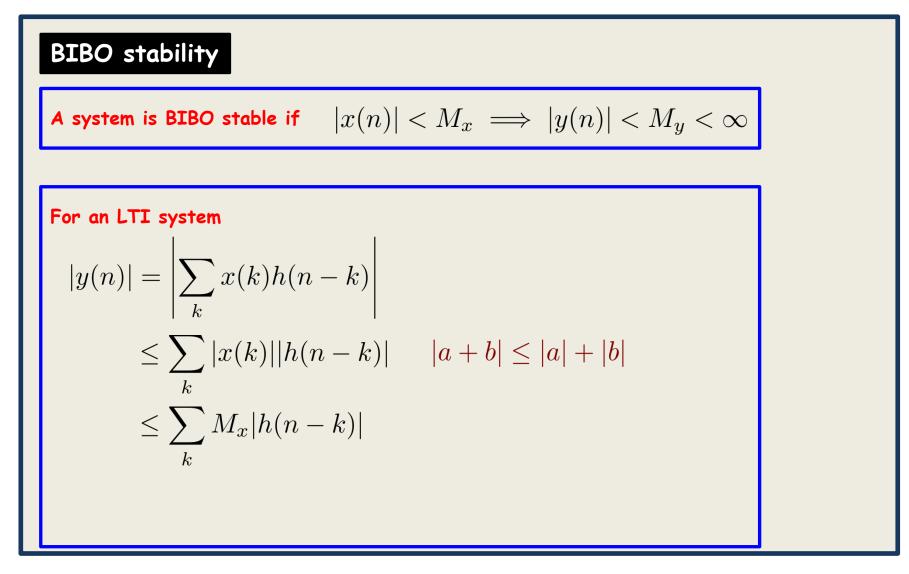


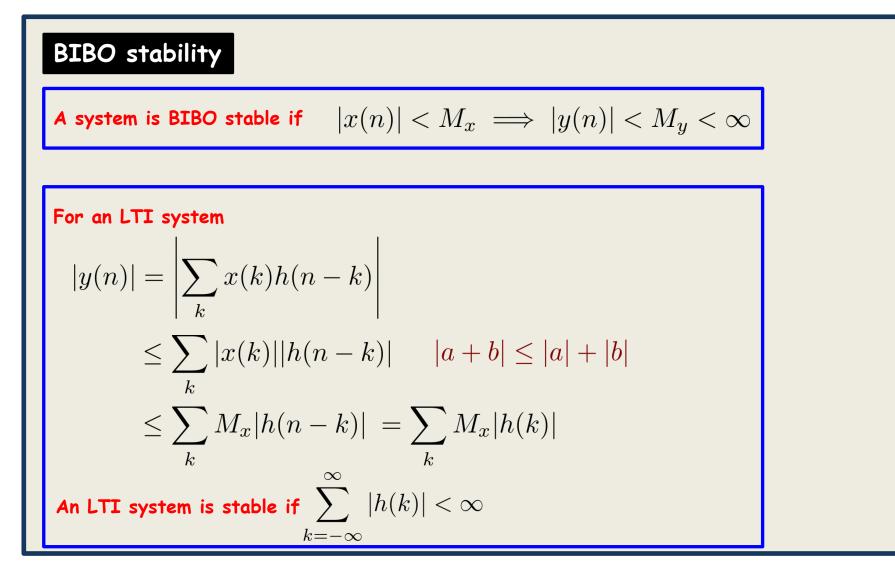












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We have seen that an LTI system is fully described by an impulse response h(n)

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We have also mentioned that difference equations are important for LTI systems

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We have also mentioned that difference equations are important for LTI systems

$$\sum_{k} a(k)y(n-k) = \sum_{\ell} b(\ell)x(n-\ell)$$

This means that every impulse response h(n) is equivalent to a difference equation We now investigate this

Relation to difference equations

$$\sum_{k} a(k)y(n-k) = \sum_{\ell} b(\ell)x(n-\ell)$$

Suppose 
$$a(k) = \begin{cases} 1 & k = 0 \\ 0 & k \neq 0 \end{cases}$$
  $b(\ell) = 0, \ \ell > L, \ell < 0$ 

Relation to difference equations

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Relation to difference equations

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The class of systems described by difference equations encompasses LTI systems with finite length impulse responses

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Relation to difference equations

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Consider now 
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$$y(1) = -a_1 y(0) + b_0 x(1) = (-a_1)^2 y(-1) + b_0 x(1) + (-a_1) b_0 x(0)$$

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$$y(2) = -a_1 y(1) + b_0 x(2)$$
  

$$= (-a_1)^3 y(-1) + b_0 x(2) + (-a_1) b_0 x(1) + (-a_1)^2 b_0 x(0)$$

Relation to difference equations

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Pattern recognition, suitably done at home, gives

n

$$y(n) = \sum_{k=0}^{n} (-a_1)^k b_0 x(n-k) + (-a_1)^{n+1} y(-1)$$

Relation to difference equations

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Convolution

Relation to difference equations

$$\sum_{k} a(k)y(n-k) = \sum_{\ell} b(\ell)x(n-\ell)$$

Consider now 
$$a(0) = 1, a(1) = a_1 \quad b(0) = b_0$$

We then get 
$$y(n) = -a_1y(n-1) + b_0x(n)$$

Pattern recognition, suitably done at home, gives

$$y(n) = \sum_{\substack{k=0\\n}}^{n} (-a_1)^k b_0 x(n-k) + (-a_1)^{n+1} y(-1)$$
$$= \sum_{\substack{k=0\\k=0}}^{n} h(k) x(n-k) + (-a_1)^{n+1} y(-1)$$
$$h(k) = (-a_1)^k b_0 u(k)$$

Infinite Impulse response (IIR)

Relation to difference equations  $\sum_k a(k)y(n-k) = \sum_\ell b(\ell)x(n-\ell)$ 

#### Infinite Impulse response (IIR)

Consider now  $a(0) = 1, a(1) = a_1 \quad b(0) = b_0$ 

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Pattern recognition, suitably done at home, gives

$$y(n) = \sum_{\substack{k=0 \ n}}^{n} (-a_1)^k b_0 x(n-k) + (-a_1)^{n+1} y(-1)$$
  
What is this?  
$$= \sum_{k=0}^{n} \frac{h(k)x(n-k)}{k} + (-a_1)^{n+1} y(-1)$$

Relation to difference equations	
$\sum_{k} a(k)y(n-k) = \sum_{\ell} b(\ell)x(n-\ell)$	
	mpulse response (IIR) $=(-a_1)^kb_0u(k)$
We then get $y(n) = -a_1y(n-1) + b_0x(n)$	
Pattern recognition, suitably done at home, gives $y(n) = \sum_{\substack{k=0 \\ n}}^{n} (-a_1)^k b_0 x(n-k) + (-a_1)^{n+1} y(-1)$ What is this? $= \sum_{k=0}^{n} h(k) x(n-k) + (-a_1)^{n+1} y(-1)$	It does not depend on ×(n)

Relation to difference equations	
$\sum_{k} a(k)y(n-k) = \sum_{\ell} b(\ell)x(n-\ell)$	
Infinite I	npulse response (IIR)
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Pattern recognition, suitably done at home, gives $n$	It does not depend on ×(n)
$y(n) = \sum_{\substack{k=0 \ n}}^{n} (-a_1)^k b_0 x(n-k) + (-a_1)^{n+1} y(-1)$ What is this? $= \sum_{k=0}^{n} \frac{h(k)}{k} x(n-k) + (-a_1)^{n+1} y(-1)$	If y(-1)≠0, we have output without any input
$-\sum_{k=0}^{n(\kappa)x(n-\kappa)+(-\mathbf{a}_1)} \mathbf{y}(-1)$	

Relation to difference equations	
$\sum_{k} a(k)y(n-k) = \sum_{\ell} b(\ell)x(n-\ell)$	
	mpulse response (IIR) $=(-a_1)^kb_0u(k)$
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Pattern recognition, suitably done at home, gives	It does not depend on ×(n)
$y(n) = \sum_{\substack{k=0 \\ n}}^{n} (-a_1)^k b_0 x(n-k) + (-a_1)^{n+1} y(-1)$ What is this?	If y(-1)≠0, we have output without any input
$= \sum_{k=0}^{n} h(k) x(n-k) + (-\mathbf{a_1})^{n+1} y(-1)$	Not Linear system Not time-invariant

Relation to difference equations	
$\sum_{k} a(k)y(n-k) = \sum_{\ell} b(\ell)x(n-\ell)$	
	Impulse response (IIR) $f(x) = (-a_1)^k b_0 u(k)$
We then get $y(n)=-a_1y(n-1)+b_0x(n)$	
If y(-1)=0, we say that the system is at rest. System is LTI	It does not depend on ×(n)
n	If y(-1)≠0, we have output without any input
$y(n) = \sum_{k=0}^{\infty} h(k)x(n-k) + (-a_1)^{n+1}y(-1)$	Not Linear system Not time-invariant

Relation to difference equations	
$\sum_{k} a(k)y(n-k) = \sum_{\ell} b(\ell)x(n-\ell)$	
	mpulse response (IIR) $=(-a_1)^kb_0u(k)$
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If y(-1)=0, we say that the system is at rest. System is LTI	It does not depend on ×(n)
If $y(-1) \neq 0$ , we say that the system is not at rest/has initial conditions. Strictly speaking: Not LTI. However, so common, so still treated within a study of LTI systems $n$	If y(-1)≠0, we have output without any input
$y(n) = \sum_{k=0}^{n} \frac{h(k)x(n-k) + (-a_1)^{n+1}y(-1)}{(-1)}$	Not Linear system Not time-invariant

Relation to difference equations	
$\sum_{k} a(k)y(n-k) = \sum_{\ell} b(\ell)x(n-\ell)$	
	mpulse response (IIR) $=(-a_1)^kb_0u(k)$
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If $y(-1)=0$ , we say that the system is at rest. System is LTI	Note: Only makes sense to assume
If $y(-1) \neq 0$ , we say that the system is not at rest/has initial conditions. Strictly speaking: Not LTI. However, so common, so	$ a_1  < 1$
still treated within a study of LTI systems $y(n) = \sum_{k=0}^n h(k) x(n-k) + (-a_1)^{n+1} y(-1)$	So, transient will fade out and "after a while it is LTI"

Relation to difference equations $\sum_k a(k)y(n-k) = \sum_\ell b(\ell)x(n-\ell)$	The class of syste difference equatio LTI systems with impulse responses	ns encompasses
Consider now $a(0) = 1, a(1) = a_1$ $b(0) = b_0$ We then get $y(n) = -a_1y(n-1) + b_0x(n)$ Infinite Impulse response (IIR) $h(k) = (-a_1)^k b_0 u(k)$		
If $y(-1)=0$ , we say that the system is at rest. System is LTI If $y(-1) \neq 0$ , we say that the system is not at rest/has initial conditions. Strictly speaking: Not LTI. However, so common, so still treated within a study of LTI systems $y(n) = \sum_{k=0}^{n} h(k)x(n-k) + (-a_1)^{n+1}y(-1)$ Note: Only makes sense to assume $ a_1  < 1$ So, transient will fade out and "after a while it is LTI"		sense to assume $ a_1  < 1$ So, transient will fade out and "after

Relation to difference equations	Every impulse response corresponds to one difference equation.	
$\sum_{k} a(k)y(n-k) = \sum_{\ell} b(\ell)x(n-\ell)$		
Consider now $a(0)=1,  a(1)=a_1$ $b(0)=b_0$ Infinite Impulse response (IIR) $h(k)=(-a_1)^kb_0u(k)$		
We then get $y(n)=-a_1y(n-1)+b_0x(n)$		
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still treated within a study of LTI systems $y(n) = \sum_{k=0}^{n} \frac{h(k)x(n-k) + (-a)}{k}$	$(1)^{n+1}y(-1)$ So, transient will fade out and "after a while it is LTI"	

#### Brief info on correlation

Not focal point of course, but highly important in signal processing

Correlation measures similarity between two signals

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Not focal point of course, but highly important in signal processing

Correlation measures similarity between two signals

Auto correlation

Cross correlation

$$r_{xx}(k) = \sum_{n = -\infty}^{\infty} x(n)x(n-k) = x(n) \star x(-n) \quad r_{yx}(k) = \sum_{n = -\infty}^{\infty} y(n)x(n-k) = y(n) \star x(-n)$$

Measures similarity between time shifted versions of the same signal

Measures similarity between time shifted versions of different signals

#### Brief info on correlation

Not focal point of course, but highly important in signal processing

Correlation measures similarity between two signals

Auto correlation

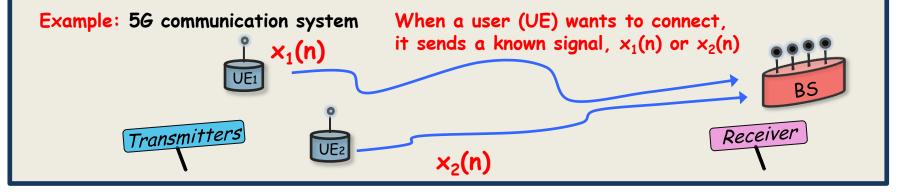
 $\infty$  $r_{xx}(k) = \sum x(n)x(n-k) = x(n) \star x(-n) \quad r_{yx}(k) = \sum y(n)x(n-k) = y(n) \star x(-n)$  $n = -\infty$ 

Measures similarity between time shifted Measures similarity between time shifted versions of the same signal

versions of different signals

Cross correlation

 $n = -\infty$ 



#### Brief info on correlation

 $\infty$ 

Not focal point of course, but highly important in signal processing

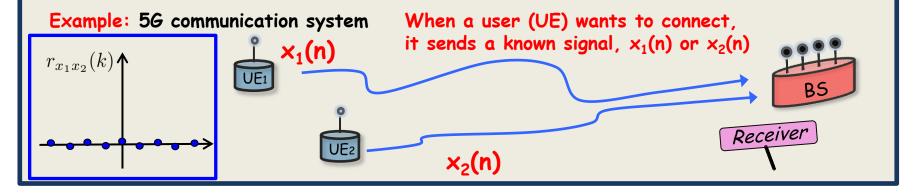
Correlation measures similarity between two signals

 $r_{xx}(k) = \sum x(n)x(n-k) = x(n) \star x(-n) \quad r_{yx}(k) = \sum y(n)x(n-k) = y(n) \star x(-n)$  $n = -\infty$ Measures similarity between time shifted Measures similarity between time shifted versions of the same signal

Auto correlation

 $n = -\infty$ versions of different signals

Cross correlation



Cross correlation between  $x_1(n)$  and  $x_2(n)$  should be small (to know who is connecting)

#### Brief info on correlation

 $\infty$ 

Not focal point of course, but highly important in signal processing

Correlation measures similarity between two signals

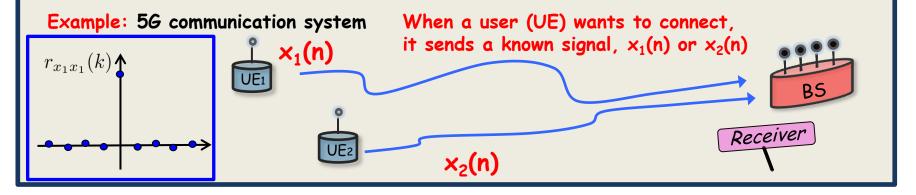
Auto correlation

 $r_{xx}(k) = \sum x(n)x(n-k) = x(n) \star x(-n) \quad r_{yx}(k) = \sum y(n)x(n-k) = y(n) \star x(-n)$  $n = -\infty$ Measures similarity between time shifted Measures similarity between time shifted versions of the same signal

versions of different signals

Cross correlation

 $n = -\infty$ 



Auto correlation of  $x_1(n)$  (and  $x_2(n)$ ) should be delta (to know when a user is connectina)

#### Brief info on correlation

Cross correlation for input and output signals

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

$$r_{yx}(k) = y(k) \star x(-k)$$

$$= x(k) \star h(k) \star x(-k)$$

$$= h(k) \star x(k) \star x(-k)$$

$$= h(k) \star x(k) \star x(-k)$$

$$= h(k) \star r_{xx}(k)$$

$$r_{yy}(k) = y(k) \star y(-k)$$

$$= x(k) \star h(k) \star x(-k) \star h(-k)$$

$$= r_{hh}(k) \star r_{xx}(k)$$