

INTRODUCTION TO EECS II
**DIGITAL
 COMMUNICATION
 SYSTEMS**

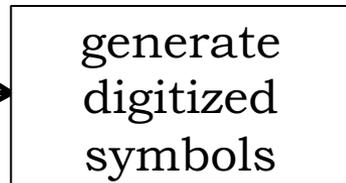
6.02 Fall 2012 Lecture #10

- Linear time-invariant (LTI) models
- Convolution

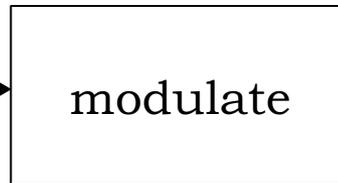
Modeling Channel Behavior

codeword
bits in

1001110101



$x[n]$



NOISY & DISTORTING ANALOG CHANNEL

ADC

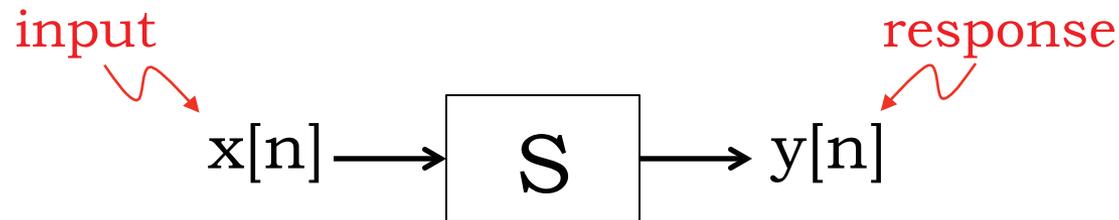
demodulate
& filter

$y[n]$

sample &
threshold

1001110101
codeword
bits out

The Baseband** Channel



A discrete-time signal such as $x[n]$ or $y[n]$ is described by an infinite sequence of values, i.e., the time index n takes values in $-\infty$ to $+\infty$. The above picture is a snapshot at a particular time n .

In the diagram above, the sequence of *output* values $y[.]$ is the *response* of system S to the *input* sequence $x[.]$

The system is **causal** if $y[k]$ depends only on $x[j]$ for $j \leq k$

**From before the modulator till after the demodulator & filter

Time Invariant Systems

Let $y[n]$ be the response of S to input $x[n]$.

If for all possible sequences $x[n]$ and integers N



then system S is said to be *time invariant* (TI). A time shift in the input sequence to S results in an identical time shift of the output sequence.

In particular, for a TI system, a shifted unit sample function $\delta[n - N]$ at the input generates an identically shifted unit sample response $h[n - N]$ at the output.

Linear Systems

Let $y_1[n]$ be the response of S to an arbitrary input $x_1[n]$ and $y_2[n]$ be the response to an arbitrary $x_2[n]$.

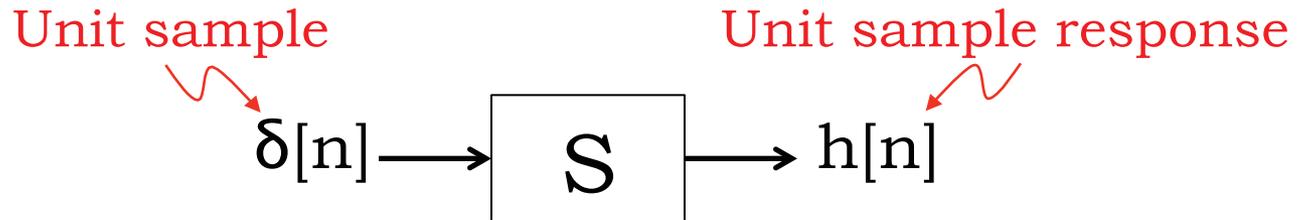
If, for arbitrary scalar coefficients a and b , we have:

$$ax_1[n] + bx_2[n] \longrightarrow \boxed{S} \longrightarrow ay_1[n] + by_2[n]$$

then system S is said to be *linear*. If the input is the weighted sum of several signals, the response is the *superposition* (i.e., *same weighted sum*) of the response to those signals.

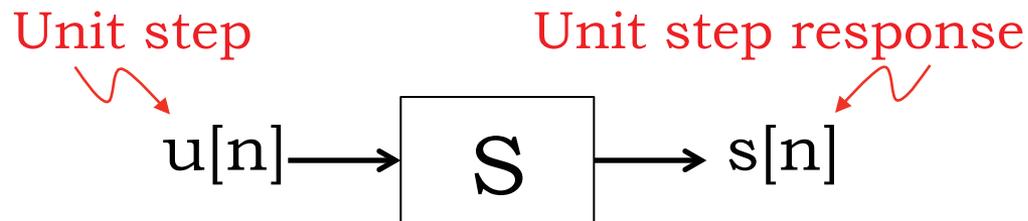
One key consequence: If the input is identically 0 for a linear system, the output must also be identically 0.

Unit Sample and Unit Step Responses

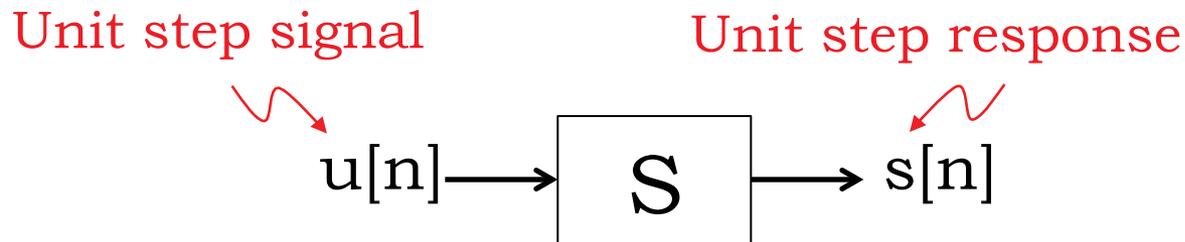
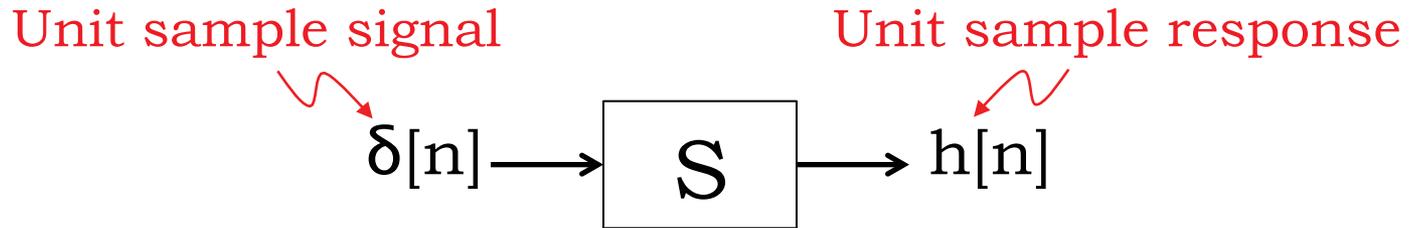


The *unit sample response* of a system S is the response of the system to the unit sample input. We will always denote the unit sample response as $h[n]$.

Similarly, the *unit step response* $s[n]$:



Relating $h[n]$ and $s[n]$ of an LTI System



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



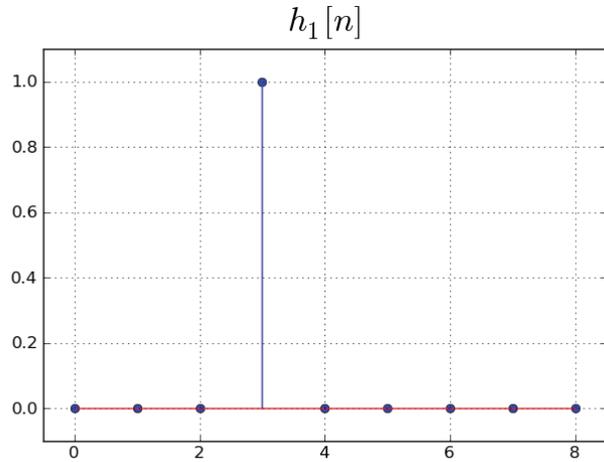
$$h[n] = s[n] - s[n-1]$$

from which it follows that

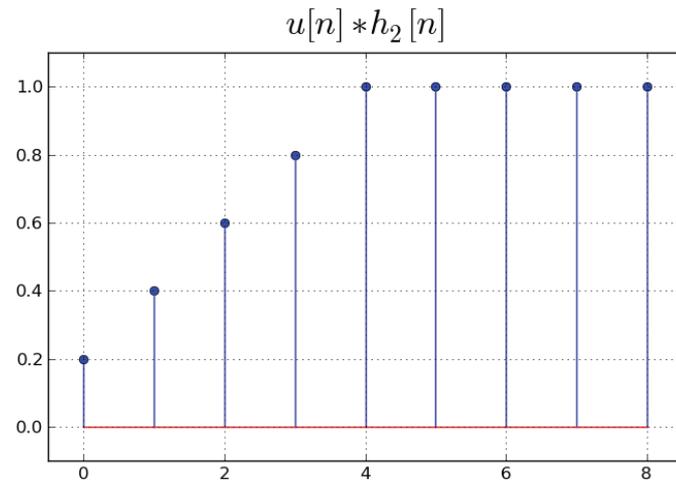
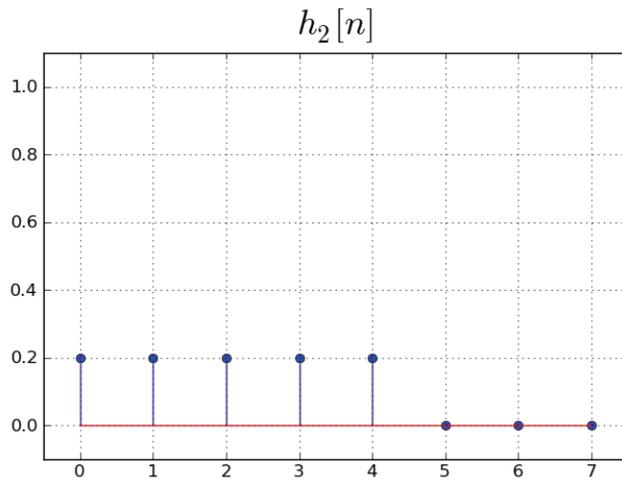
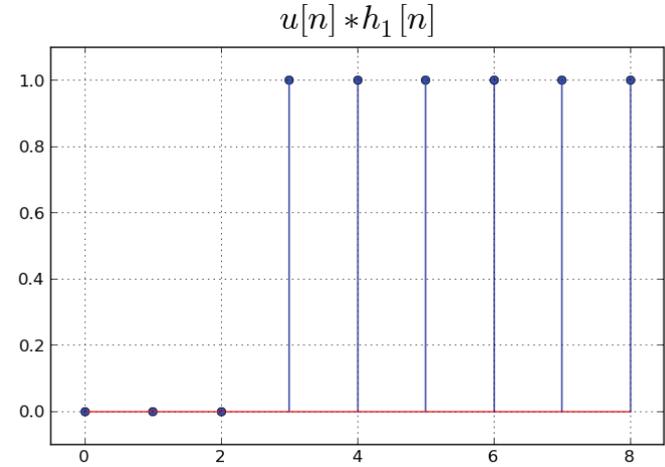
$$s[n] = \sum_{k=-\infty}^n h[k]$$

(assuming $s[-\infty] = 0$, e.g., a **causal** LTI system; more generally, a “**right-sided**” unit sample response)

$h[n]$

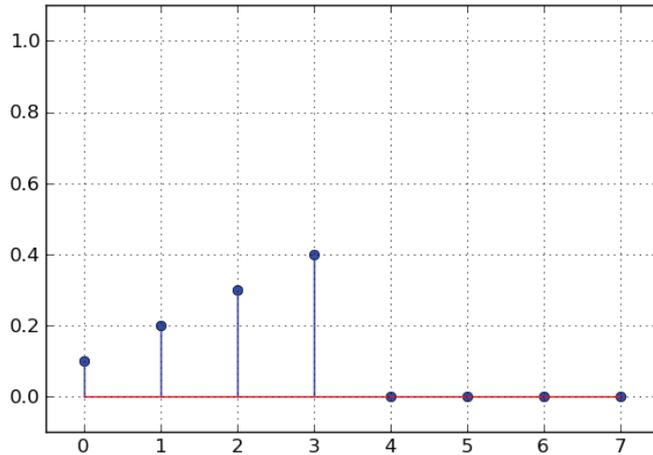


$s[n]$



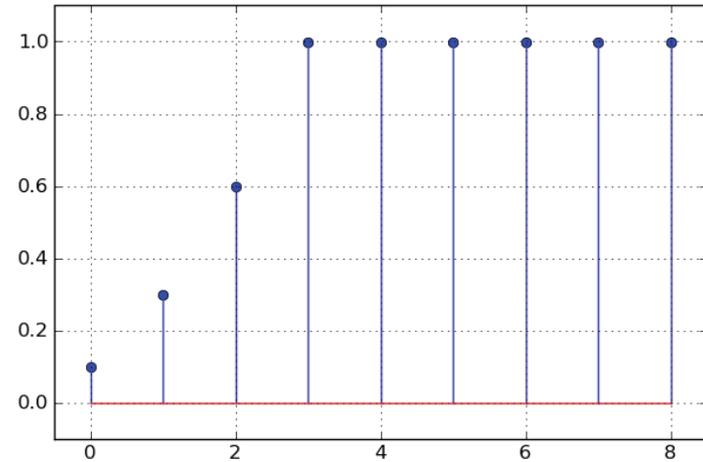
$h[n]$

$h_3[n]$

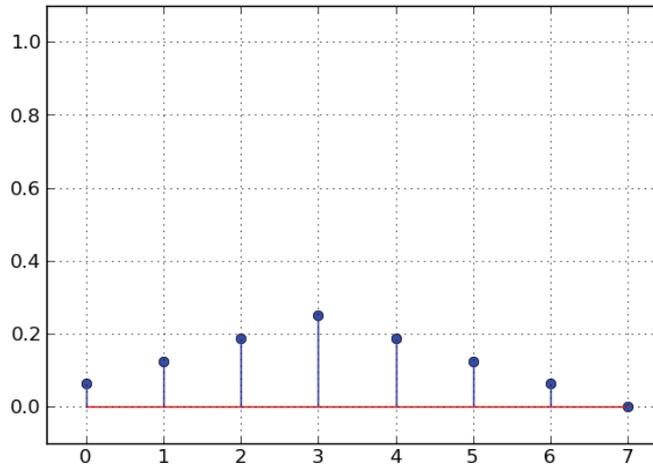


$s[n]$

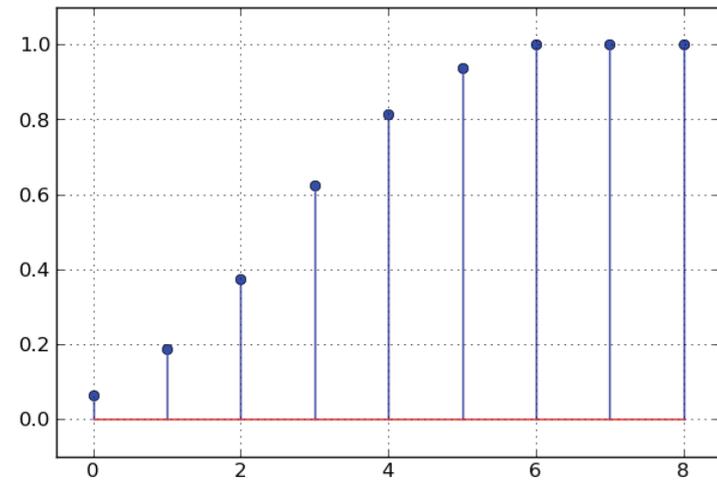
$u[n] * h_3[n]$



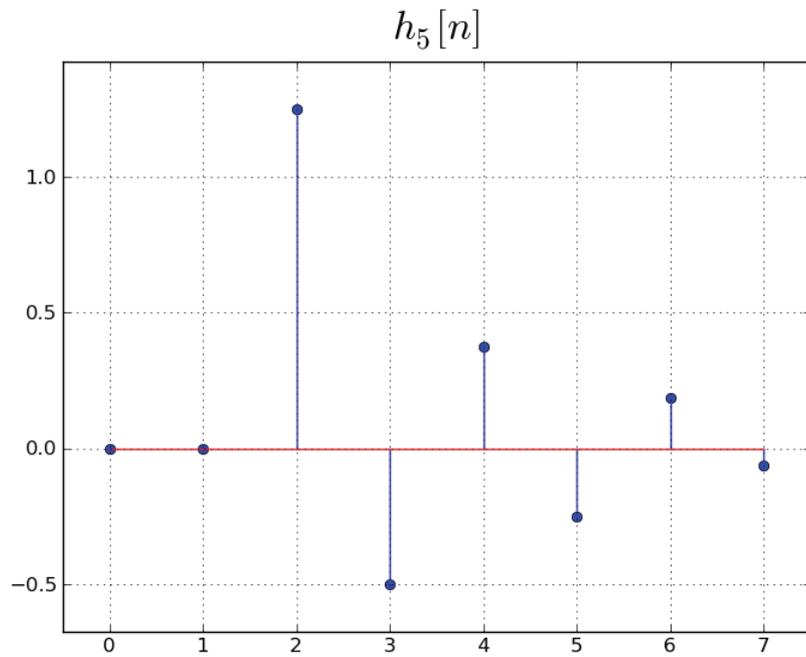
$h_4[n]$



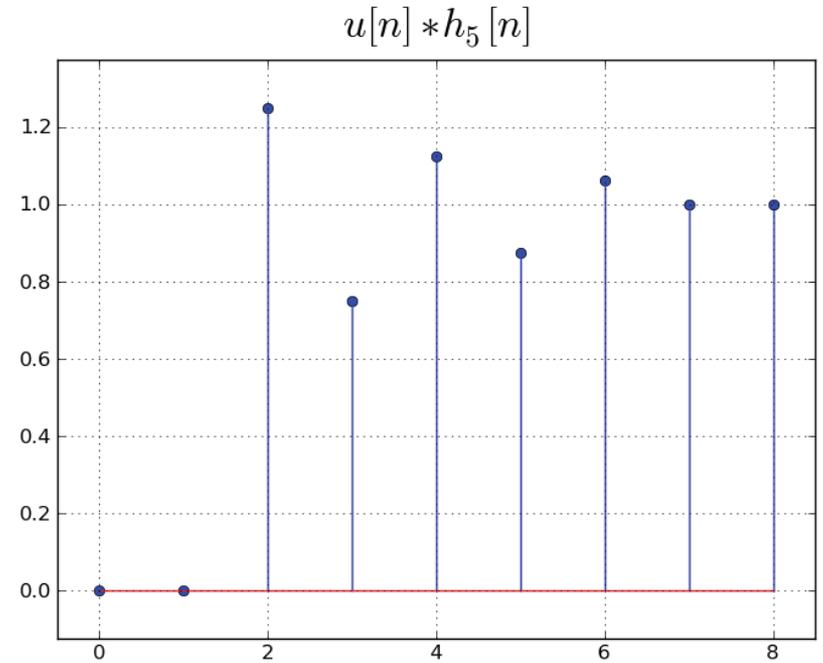
$u[n] * h_4[n]$



$h[n]$



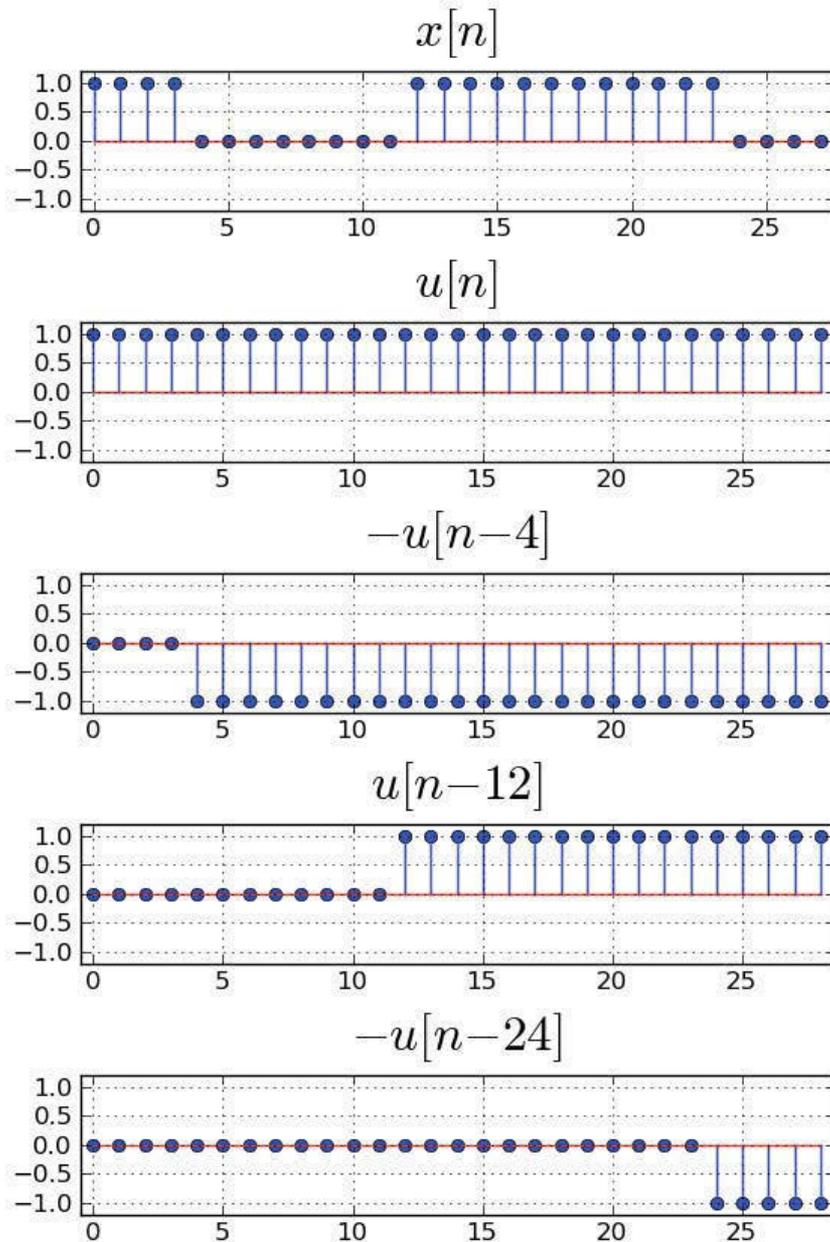
$s[n]$



Unit Step Decomposition

“Rectangular-wave” digital signaling waveforms, of the sort we have been considering, are easily decomposed into **time-shifted, scaled unit steps** --- each transition corresponds to another shifted, scaled unit step.

e.g., if $x[n]$ is the transmission of 1001110 using 4 samples/bit:



$$\begin{aligned}
 x[n] &= u[n] \\
 &\quad - u[n-4] \\
 &\quad + u[n-12] \\
 &\quad - u[n-24]
 \end{aligned}$$

... so the corresponding response is

$$x[n]$$

$$= u[n]$$

$$- u[n - 4]$$

$$+ u[n - 12]$$

$$- u[n - 24]$$



$$y[n]$$

$$= s[n]$$

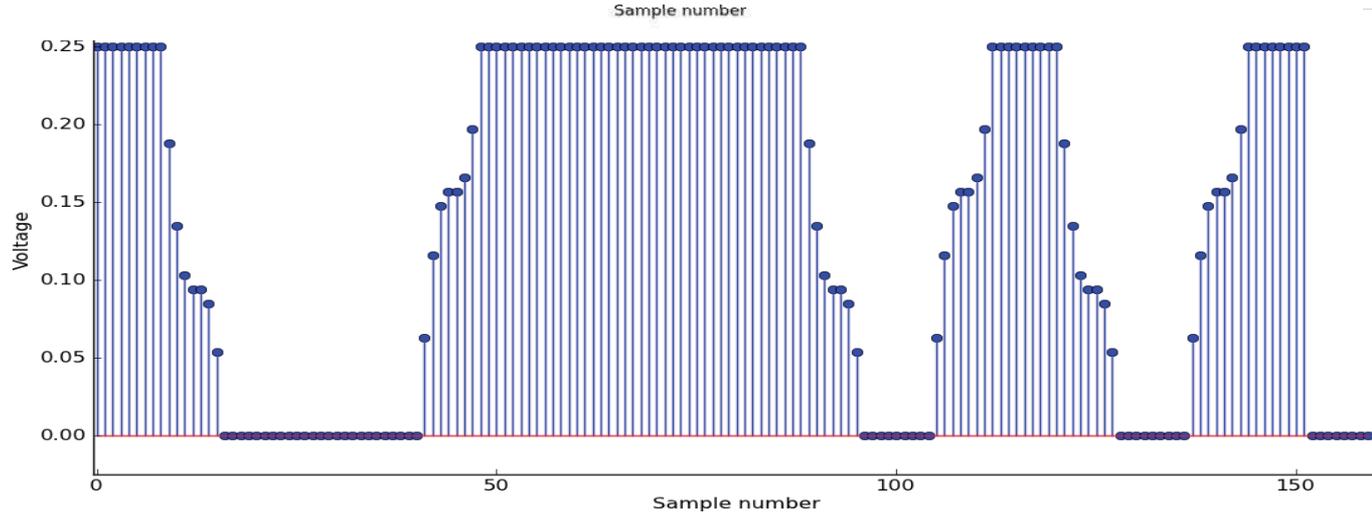
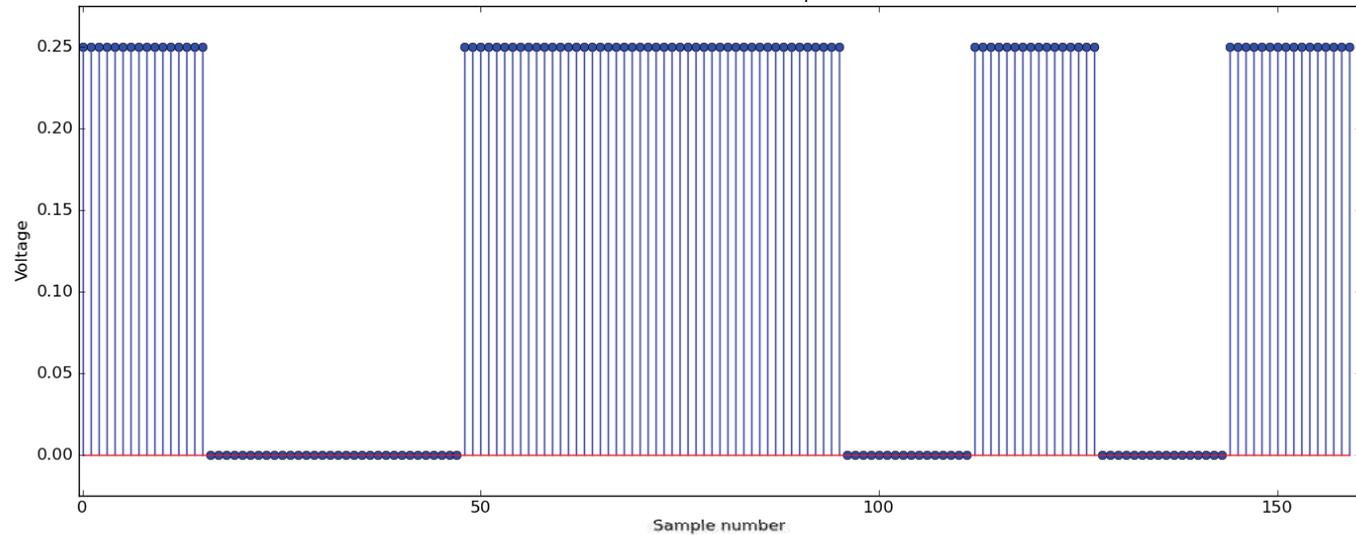
$$- s[n - 4]$$

$$+ s[n - 12]$$

$$- s[n - 24]$$

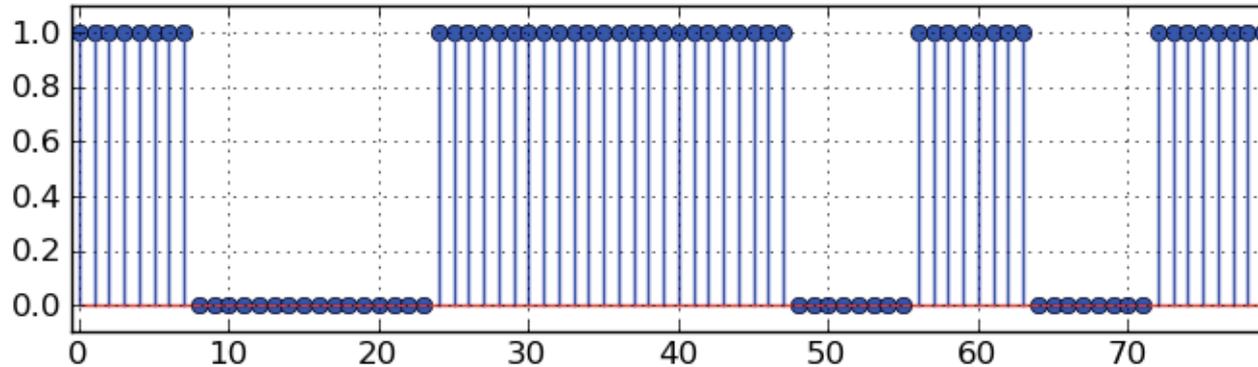
Note how we have invoked linearity and time invariance!

Example

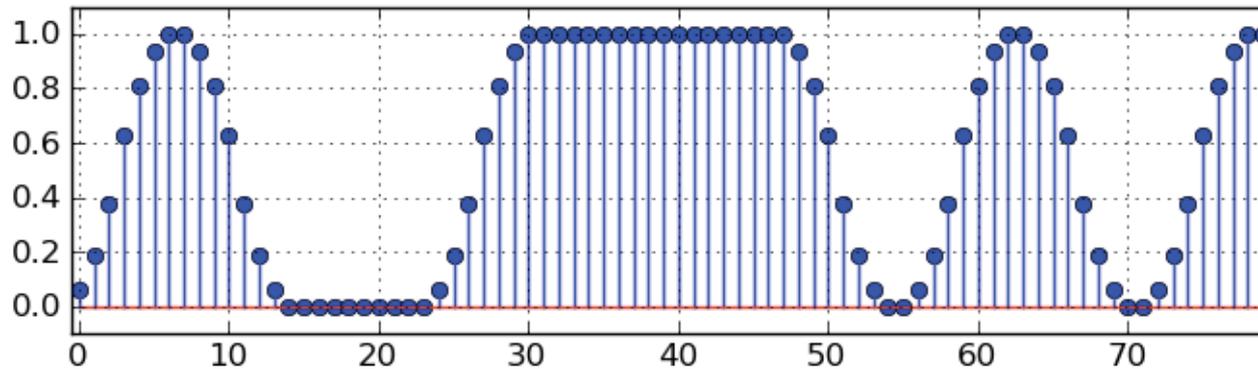


Transmission Over a Channel

$x[n]$ at 8 samples/bit



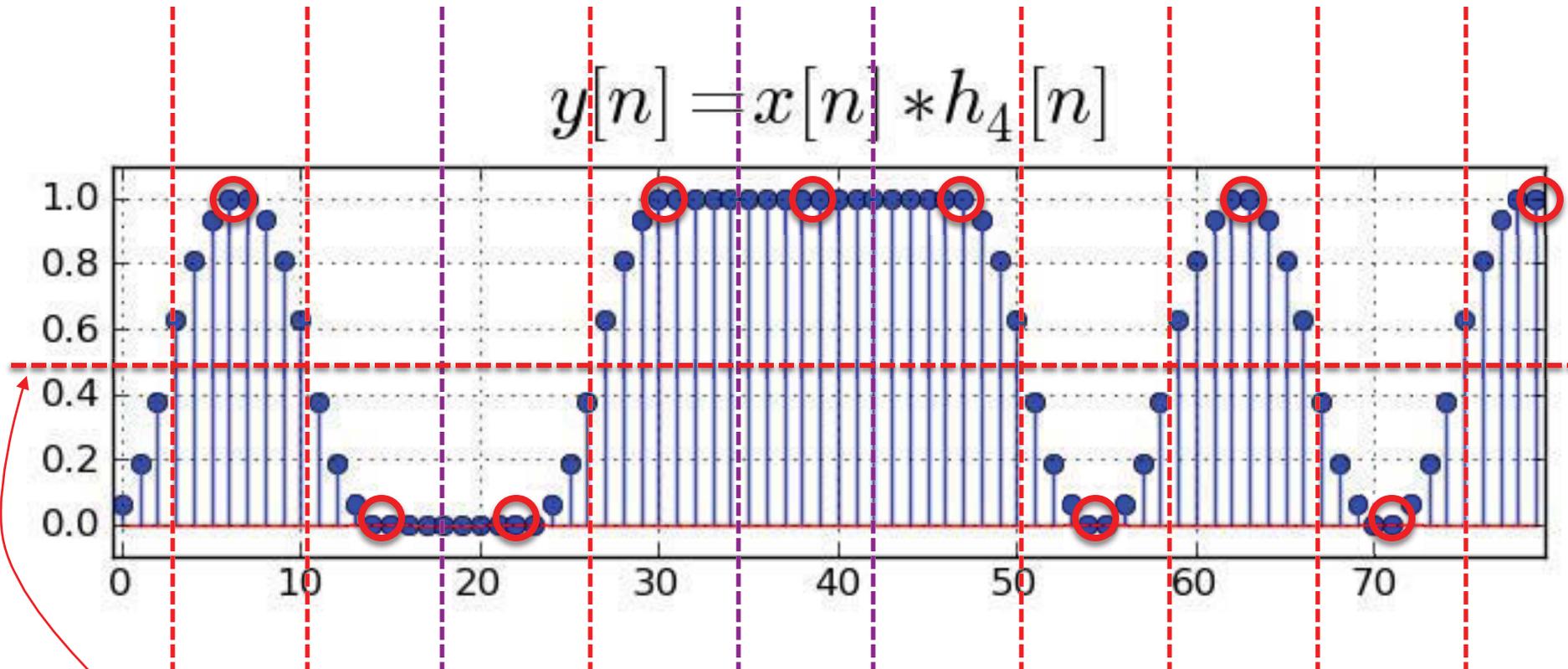
$$y[n] = x[n] * h_4[n]$$



Ignore this notation for now, will explain shortly

Receiving the Response

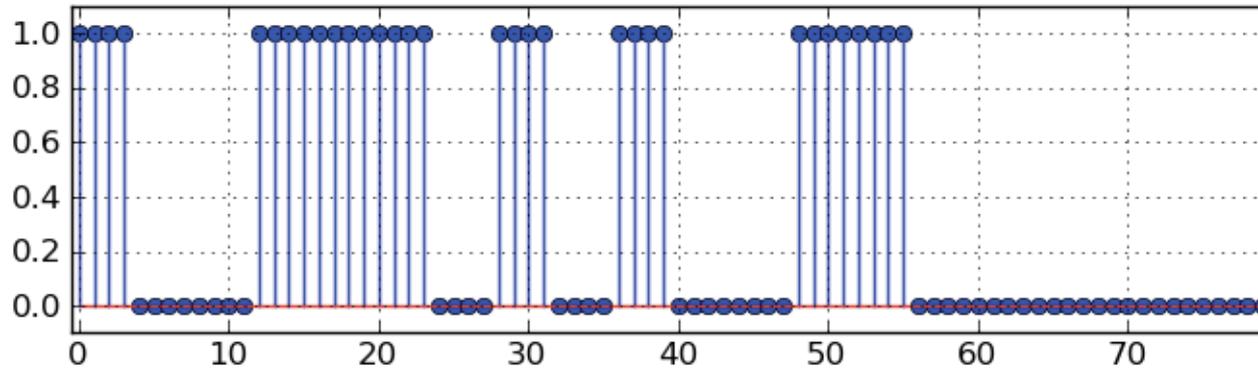
$$y[n] = x[n] * h_4[n]$$



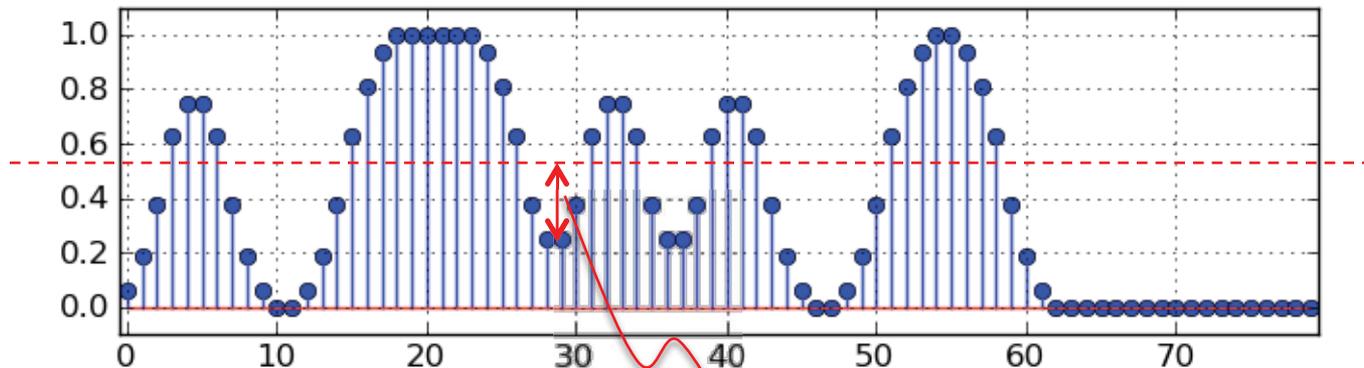
Digitization threshold = 0.5V

Faster Transmission

$x[n]$ at 4 samples/bit



$y[n] = x[n] * h_4[n]$



Unit Sample Decomposition

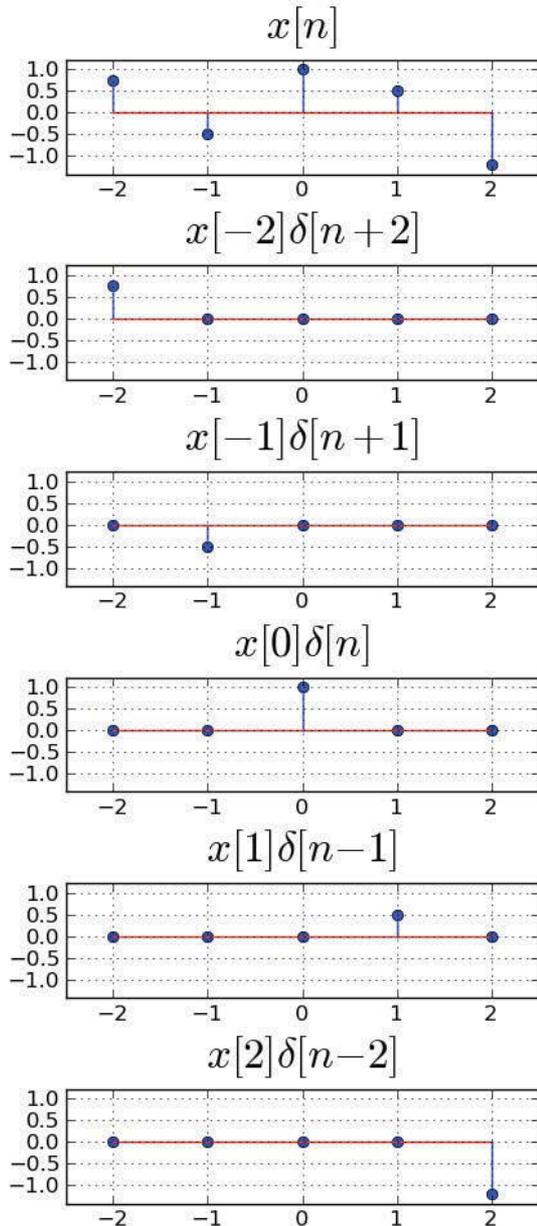
A discrete-time signal can be decomposed into a sum of time-shifted, scaled unit samples.

Example: in the figure, $x[n]$ is the sum of $x[-2]\delta[n+2] + x[-1]\delta[n+1] + \dots + x[2]\delta[n-2]$.

In general:

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

For any particular index, only one term of this sum is non-zero



Modeling LTI Systems

If system S is both linear and time-invariant (LTI), then we can use the unit sample response to predict the response to *any* input waveform $x[n]$:

Sum of shifted, scaled unit samples

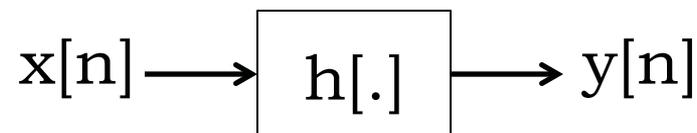
$$x[n] = \sum_{k=-\infty}^{\infty} x[k] \delta[n-k] \longrightarrow \boxed{S}$$

Sum of shifted, scaled responses

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

CONVOLUTION SUM

Indeed, the unit sample response $h[n]$ completely characterizes the LTI system S , so you often see



Convolution

Evaluating the convolution sum

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

for all n defines the output signal y in terms of the input x and unit-sample response h . Some constraints are needed to ensure this infinite sum is well behaved, i.e., doesn't "blow up" --- we'll discuss this later.

We use $*$ to denote convolution, and write $y=x*h$. We can then write the value of y at time n , which is given by the above sum, as $y[n] = (x * h)[n]$. We could perhaps even write $y[n] = x * h[n]$

Convolution

Evaluating the convolution sum

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

for all n defines the output signal y in terms of the input x and unit-sample response h . Some constraints are needed to ensure this infinite sum is well behaved, i.e., doesn't "blow up" --- we'll discuss this later.

We use $*$ to denote convolution, and write $y=x*h$. We can thus write the value of y at time n , which is given by the above sum, as $y[n] = (x * h)[n]$

Instead you'll find people writing $y[n] = x[n] * h[n]$, where the poor index n is doing double or triple duty. This is **awful** notation, but a super-majority of engineering professors (including at MIT) will inflict it on their students.

Don't stand for it!

Properties of Convolution

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

The second equality above establishes that convolution is **commutative**:

$$x * h = h * x$$

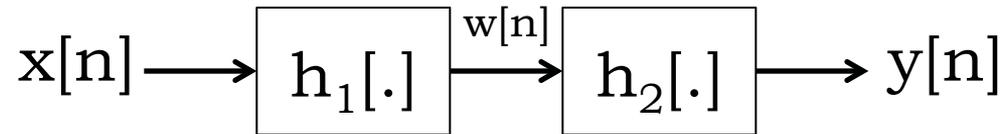
Convolution is **associative**:

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

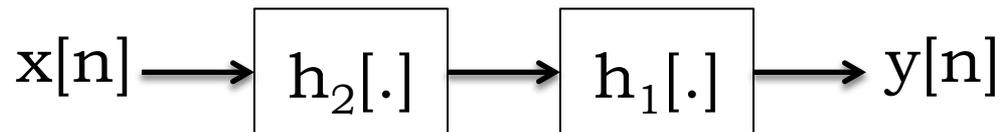
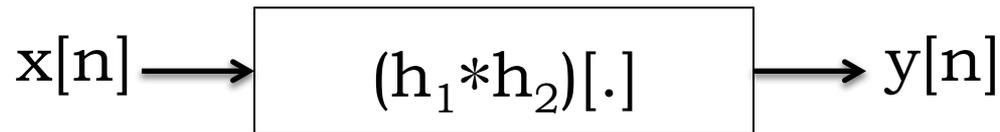
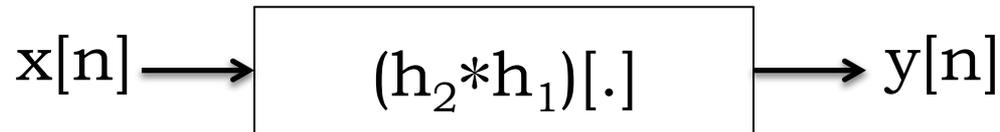
Convolution is **distributive**:

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

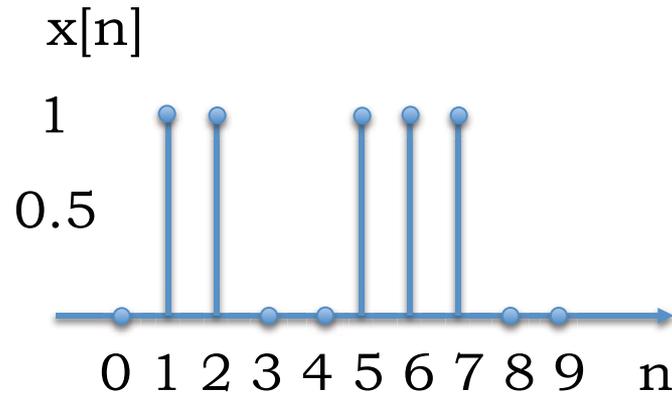
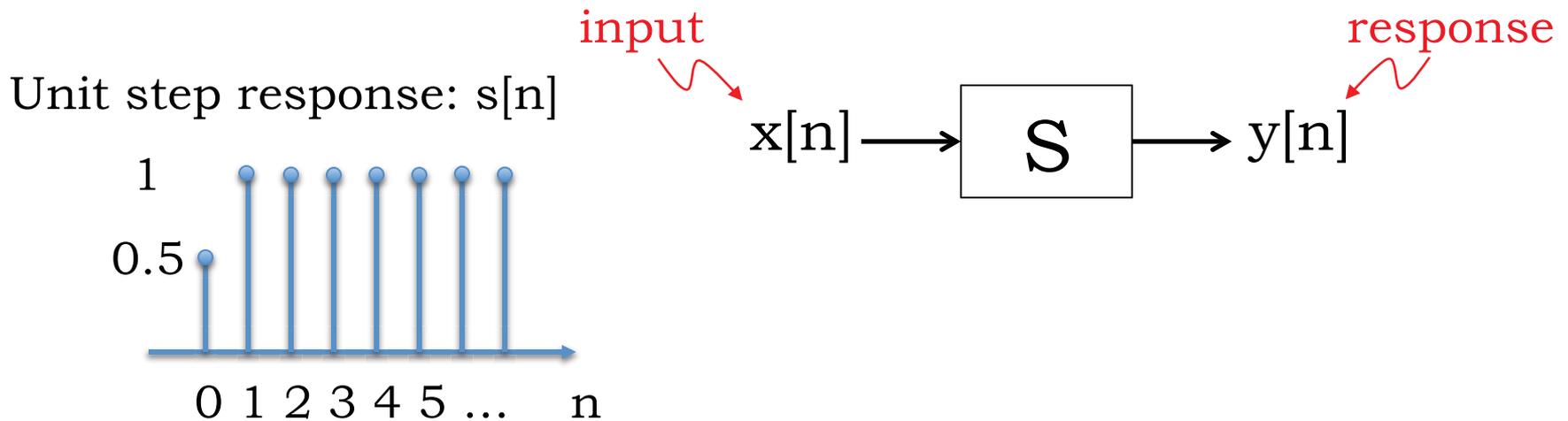
Series Interconnection of LTI Systems



$$y = h_2 * w = h_2 * (h_1 * x) = (h_2 * h_1) * x$$



Spot Quiz



Find $y[n]$:

1. Write $x[n]$ as a function of unit steps
2. Write $y[n]$ as a function of unit step responses
3. Draw $y[n]$

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Fall 2012

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