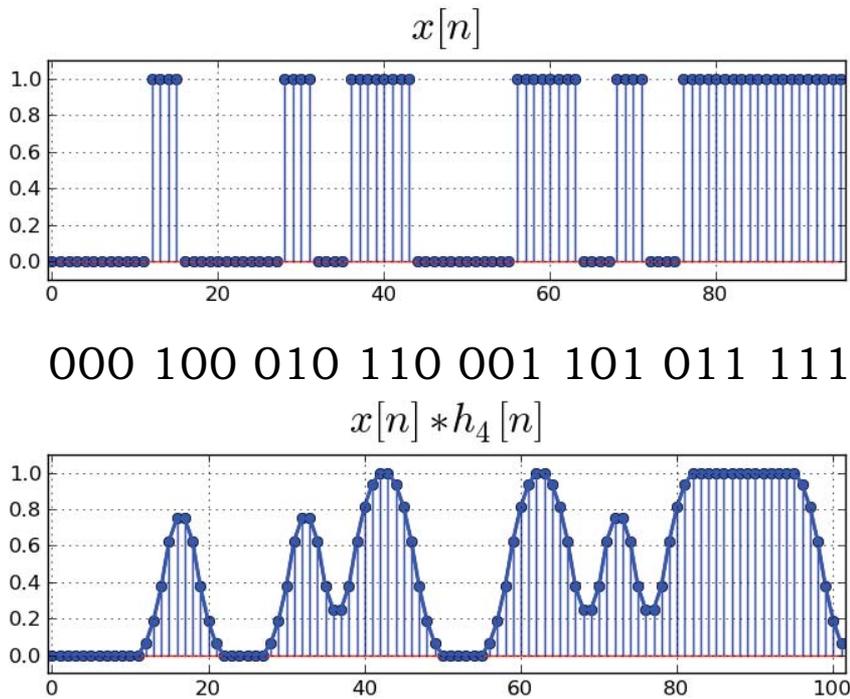


INTRODUCTION TO EECS II  
**DIGITAL  
 COMMUNICATION  
 SYSTEMS**

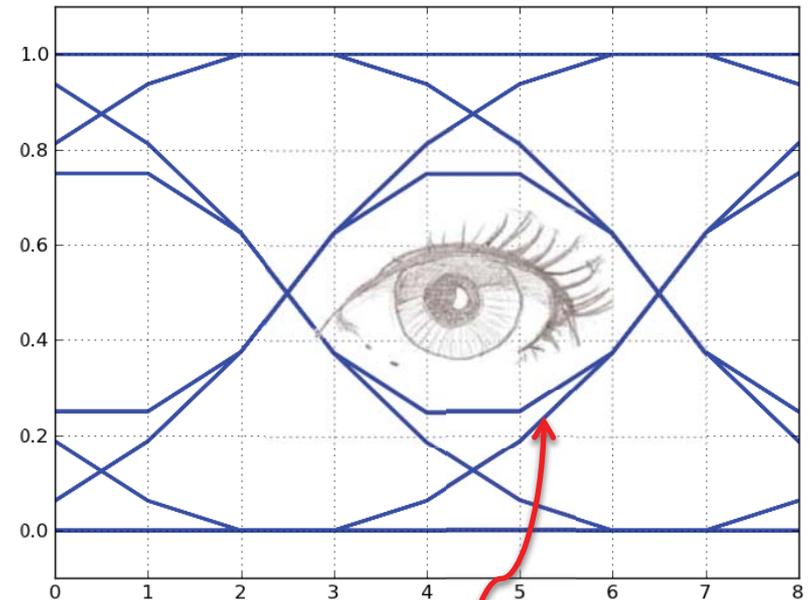
# 6.02 Fall 2012 Lecture #11

- Eye diagrams
- Alternative ways to look at convolution

# Eye Diagrams



Eye diagram:  $h_4[n]$ , 4 samples/bit

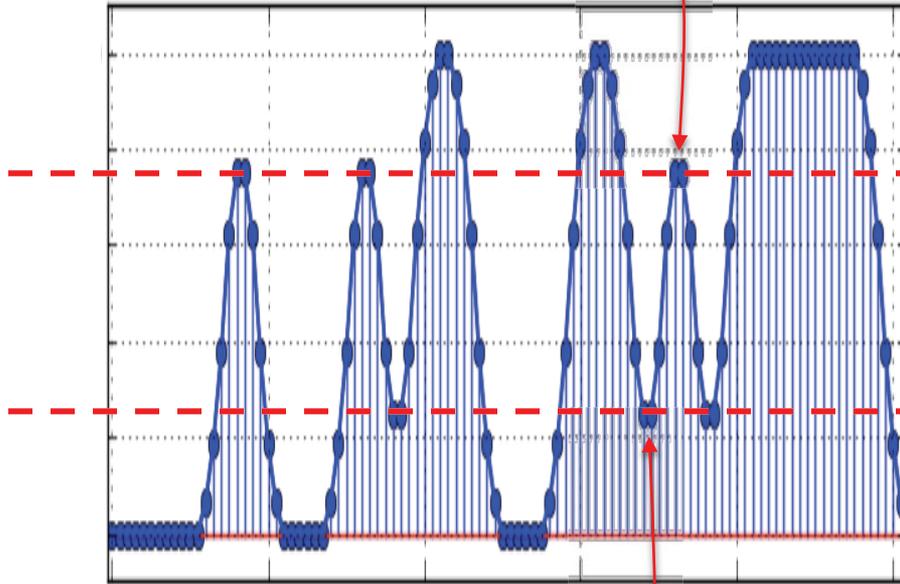


Eye diagrams make it easy to find the worst-case signaling conditions at the receiving end.

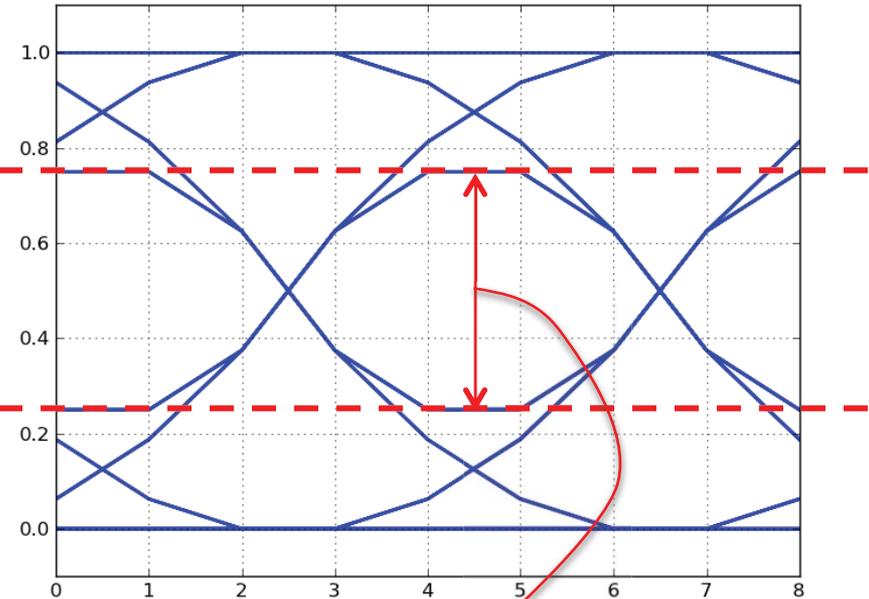
These are overlaid two-bit-slot segments of step responses, plotted without the 'stems' of the stem plot on the left

# “Width” of Eye

Worst-case “1”



Worst-case “0”



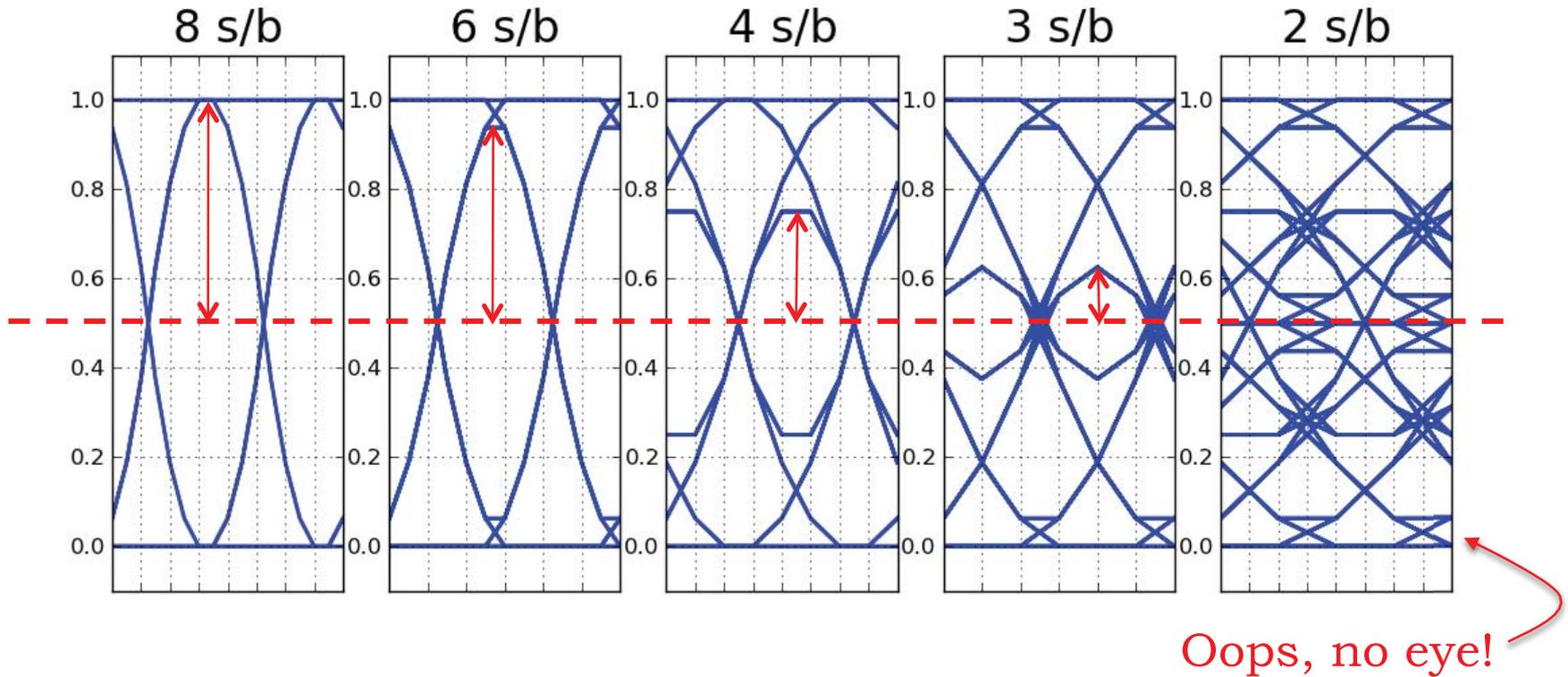
“width” of eye  
(as in “eye wide open”)

To maximize noise margins:

Pick the best sample point → widest point in the eye

Pick the best digitization threshold → half-way across width

# Choosing Samples/Bit

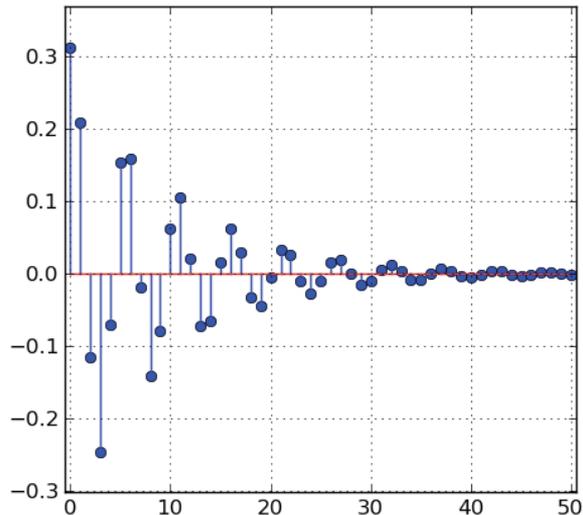


Given  $h[n]$ , you can use the eye diagram to pick the number of samples transmitted for each bit ( $N$ ):

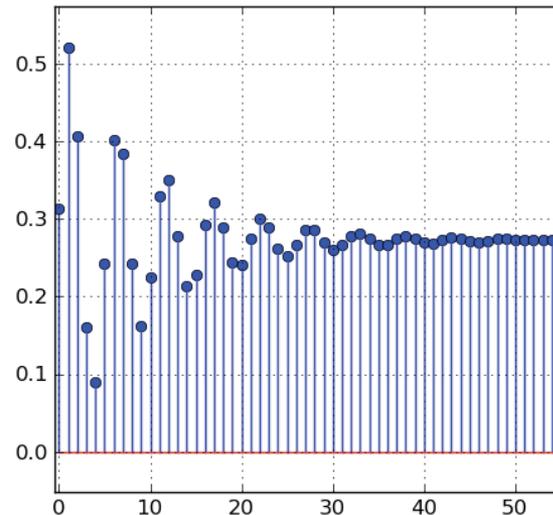
Reduce  $N$  until you reach the noise margin you feel is the minimum acceptable value.

# Example: “ringing” channel

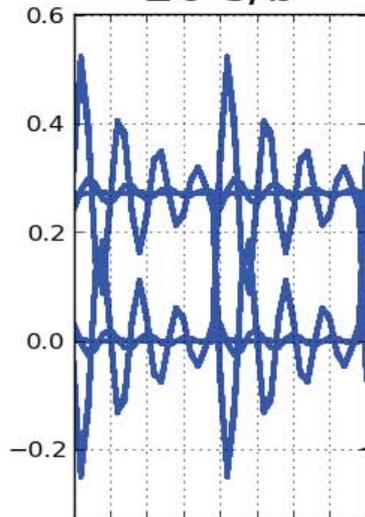
$$h_{RING}[n]$$



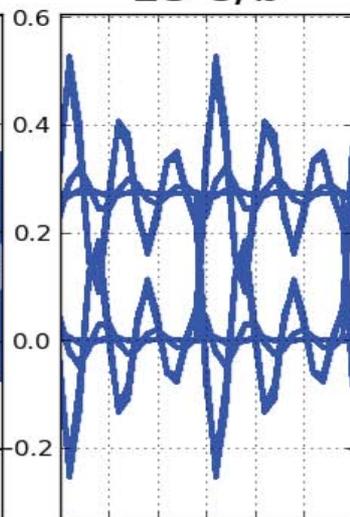
$$s_{RING}[n]$$



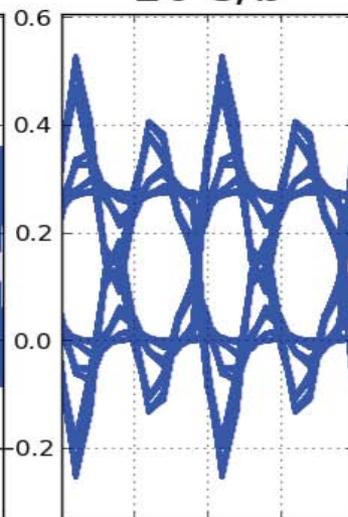
20 s/b



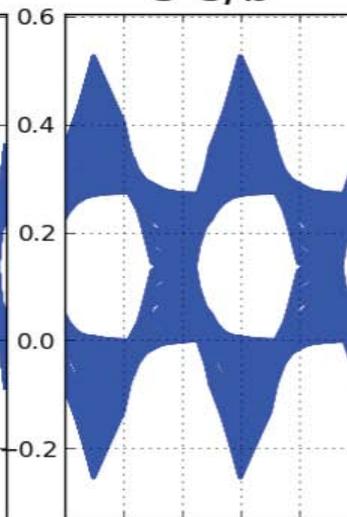
15 s/b



10 s/b



5 s/b



# Constructing the Eye Diagram

(no need to wade through all this unless you really want to!)

1. Generate an input bit sequence pattern that contains all possible combinations of  $B$  bits (e.g.,  $B=3$  or  $4$ ), so a sequence of  $2^B B$  bits. (Otherwise, a random sequence of comparable length is fine.)
2. Transmit the corresponding  $x[n]$  over the channel ( $2^B B N$  samples, if there are  $N$  samples/bit)
3. Instead of one long plot of  $y[n]$ , plot the response as an *eye diagram*:
  - a. break the plot up into short segments, each containing  $KN$  samples, starting at sample  $0, KN, 2KN, 3KN, \dots$  (e.g.,  $K=2$  or  $3$ )
  - b. plot all the short segments on top of each other

# Back To Convolution

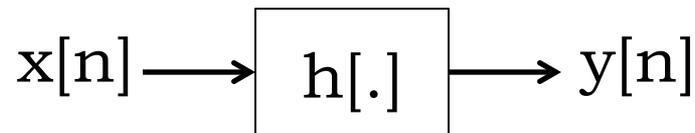
From last lecture: If system S is both linear and time-invariant (LTI), then we can use the unit sample response  $h[n]$  to predict the response to *any* input waveform  $x[n]$ :

Sum of shifted, scaled unit sample functions Sum of shifted, scaled unit sample responses, with the same scale factors

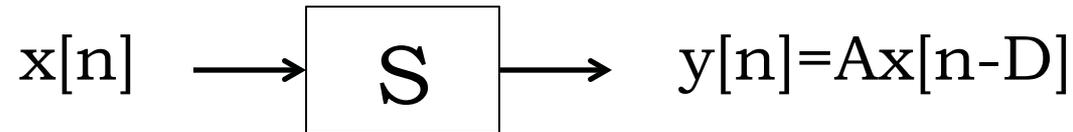
$$x[n] = \sum_{k=-\infty}^{\infty} x[k] \delta[n-k] \longrightarrow \boxed{S} \longrightarrow y[n] = \sum_{k=-\infty}^{\infty} x[k] h[n-k]$$

CONVOLUTION SUM

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



# Unit Sample Response of a Scale-&-Delay System



If  $S$  is a system that scales the input by  $A$  and delays it by  $D$  time steps (negative 'delay'  $D =$  advance), is the system

time-invariant? **Yes!**

linear? **Yes!**

Unit sample response is  $h[n] = A\delta[n-D]$

General unit sample response

$$h[n] = \dots + h[-1] \delta[n+1] + h[0] \delta[n] + h[1] \delta[n-1] + \dots$$

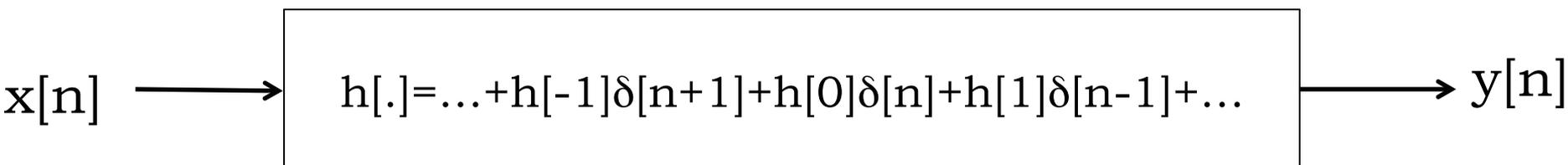
for an LTI system can be thought of as resulting from many scale-&-delays in parallel

# A Complementary View of Convolution

So instead of the picture:

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

we can consider the picture:



from which we get

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

(To those who have an eye for these things, my apologies

# (side by side)

$$y[n] =$$

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

Input term  $x[0]$  at time 0 launches scaled unit sample response  $x[0]h[n]$  at output

Input term  $x[k]$  at time  $k$  launches scaled shifted unit sample response  $x[k]h[n-k]$  at output

Unit sample response term  $h[0]$  at time 0 contributes scaled input  $h[0]x[n]$  to output

Unit sample response term  $h[m]$  at time  $m$  contributes scaled shifted input  $h[m]x[n-m]$  to output

# To Convolve (but not to “Convolute”!)

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

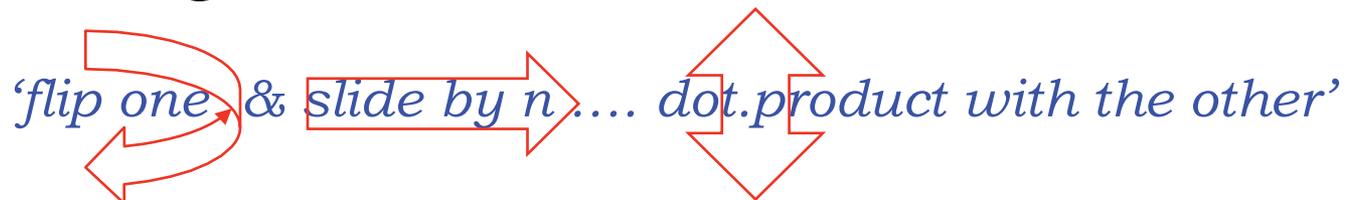
A simple graphical implementation:

Plot  $x[.]$  and  $h[.]$  as a function of the dummy index (k or m above)

**Flip** (i.e., reverse) one signal in time,  
**slide** it right **by n** (slide left if n is -ve), take the  
**dot.product** with the other.

This yields the value of the convolution at the single time n.

*‘flip one & slide by n ... dot.product with the other’*



# Example

- From the unit **sample** response  $h[n]$  to the unit **step** response

$$s[n] = (h * u)[n]$$

- **Flip**  $u[k]$  to get  $u[-k]$
- **Slide**  $u[-k]$   $n$  steps to right (i.e., delay  $u[-k]$ ) to get  $u[n-k]$ , place over  $h[k]$
- **Dot product of**  $h[k]$  and  $u[n-k]$  wrt  $k$ :

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

# Channels as LTI Systems

Many transmission channels can be effectively modeled as LTI systems. When modeling transmissions, there are few simplifications we can make:

- We'll call the time transmissions start  $t=0$ ; the signal before the start is 0. So  $x[m] = 0$  for  $m < 0$ .
- Real-world channels are **causal**: the output at any time depends on values of the input at only the present and past times. So  $h[m] = 0$  for  $m < 0$ .

These two observations allow us to rework the convolution sum when it's used to describe transmission channels:

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

  
start at  $t=0$       causal       $j=n-k$

# 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**:

$$x * h = h * x$$

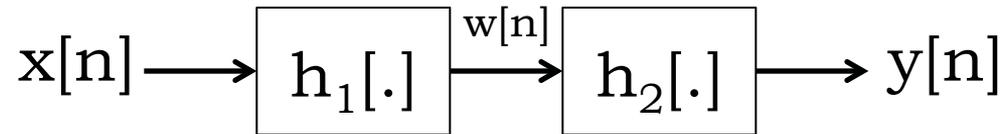
Convolution is **associative**:

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

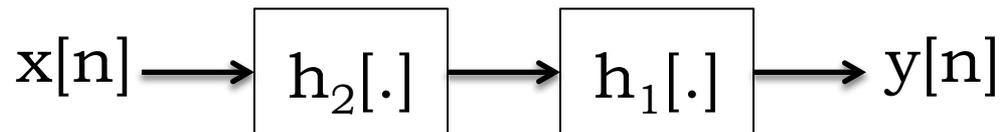
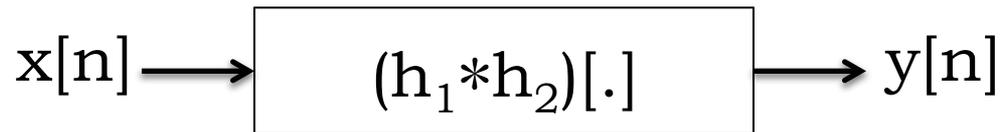
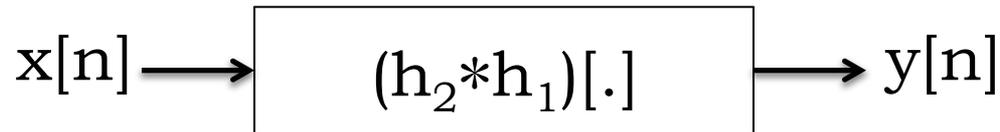
Convolution is **distributive**:

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

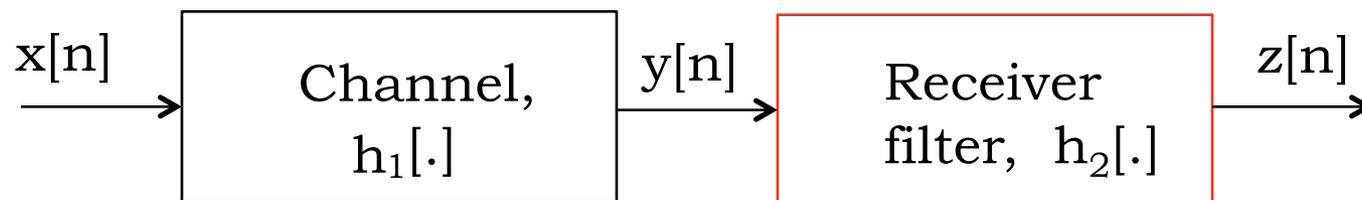
# Series Interconnection of LTI Systems



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



# “Deconvolving” Output of Echo Channel



Suppose channel is LTI with

$$h_1[n] = \delta[n] + 0.8\delta[n-1]$$

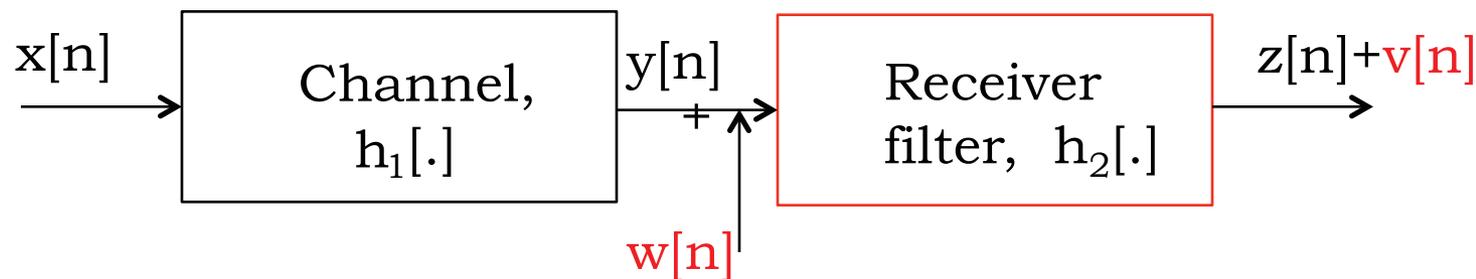
Find  $h_2[n]$  such that  $z[n] = x[n]$



$$(h_2 * h_1)[n] = \delta[n]$$

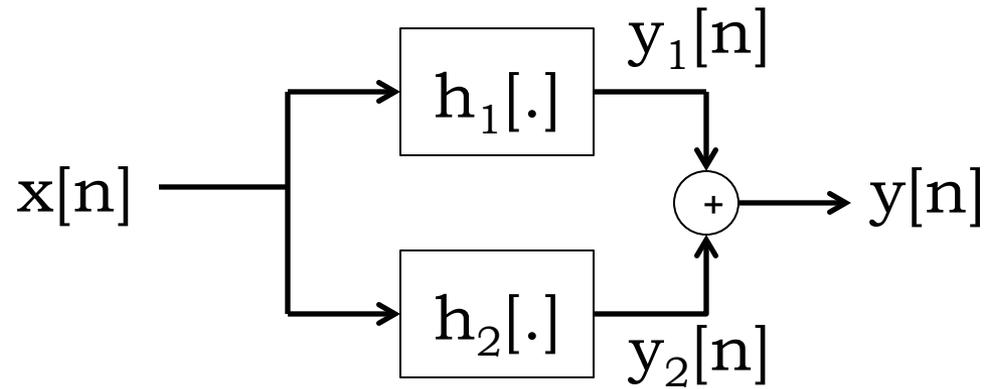
Good exercise in applying  
Flip/Slide/Dot.Product

# “Deconvolving” Output of Channel with Echo

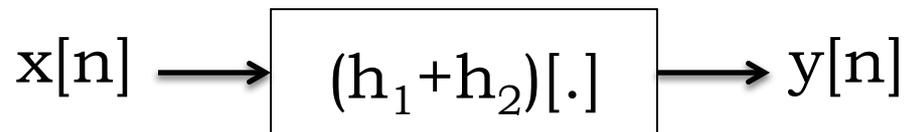


Even if channel was well modeled as LTI and  $h_1[n]$  was known, **noise** on the channel can greatly degrade the result, so this is usually not practical.

# Parallel Interconnection of LTI Systems



$$y = y_1 + y_2 = (h_1 * x) + (h_2 * x) = (h_1 + h_2) * x$$



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Fall 2012

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