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# EE102

## Lecture 7

# EE102 Announcements

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- Syllabus link is [tiny.cc/ucla102](https://tiny.cc/ucla102)
- CCLE difficulties, please email [help@seas.ucla.edu](mailto:help@seas.ucla.edu)
- **Third Homework due this Friday**
- **Note: We do not have lectures on exam days.**

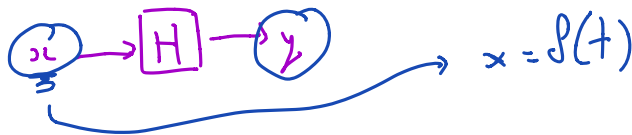
**Slide Credits:** This lecture adapted from Prof. Jonathan Kao (UCLA) and recursive credits apply to Prof. Cabric (UCLA), Prof. Yu (Carnegie Mellon), Prof. Pauly (Stanford), and Prof. Fragouli (UCLA)

# Review of Last Lecture

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Last Lecture Introduced a few concepts:

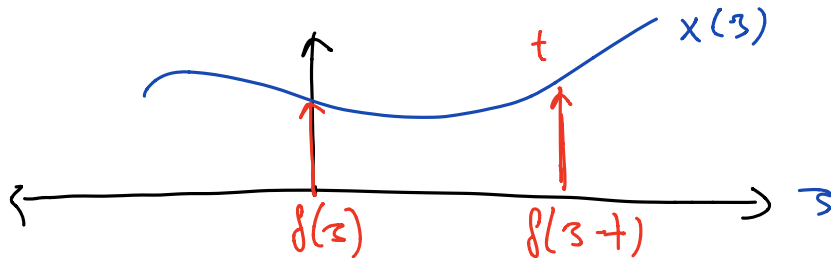
- CYU: How is the Impulse Response Defined?



- CYU: Why is the Impulse Response Useful?

IR lets you predict the response of an LTI system, not just to the impulse (of course), but to Any input

# Derivation of this fact



① Calculate  $x(0)$  using  $\delta$ 's:  $x(z)\delta(z) = \underbrace{x(0)\delta(z)}_{\text{Integrate this, you get } x(0).}$

② Calculate  $x(t)$  using  $\delta$ 's:  $x(z)\delta(z-t) = \underbrace{x(t)\delta(z-t)}_{\text{Integrate to get } x(t)}$

# The Convolution Integral

$$\begin{aligned} \int_{-\infty}^{\infty} x(z) \delta(z-t) dz &= \int_{-\infty}^{\infty} \underbrace{x(t)} \delta(z-t) dz \\ &= x(t) \underbrace{\int_{-\infty}^{\infty} \delta(z-t) dz}_{=1} \\ &= x(t) \end{aligned}$$

## // Convolution Integral

$$x(t) = \int_{-\infty}^{\infty} x(s) \delta(s-t) ds$$

$$= \int_{-\infty}^{\infty} x(s) \delta(t-s) ds$$

### Canonical form of Convolution Integral.

# The Convolution Integral (Cont'd)

$$h(t) \triangleq H(\delta(t))$$

① If we have  $h(t)$   
then given any  $x(t)$  we  
can get  $y(t)$



② Given a system  $H$  and  $x$ , predict  $y$

// Goal

$$\textcircled{3} \quad y(t) = H(x(t))$$

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

Linearity  
Time Invariance

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

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

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

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

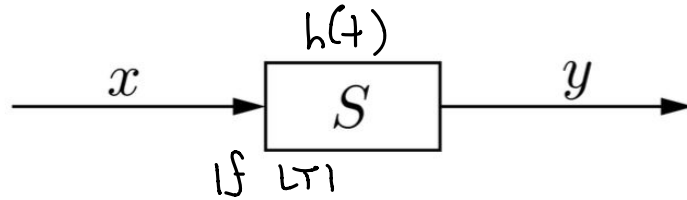
$$y(t) = x(t) \star h(t)$$

$\equiv$   
convolution

# Convolution is what adds structure to the Black Box

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A system transforms an *input signal*,  $x(t)$ , into an output system,  $y(t)$ .



Then, convolution is the operation that maps  $x$  to  $y$ ,

$$y = x \star h$$

# Why does this work? (The Convolution Integral)

Same derivation as last lecture - shows up in other classes, even grad classes

without loss of generality  $x \rightarrow \boxed{H} \rightarrow y$

$$y(t) = H[x(t)]$$

$$y(t) = H\left[\int_{-\infty}^{\infty} x(\tau) \delta(t-\tau) d\tau\right]$$

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

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

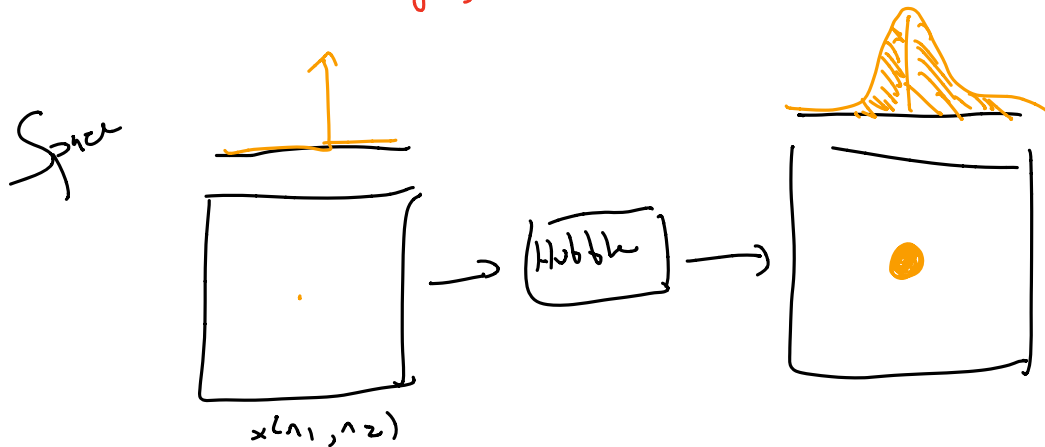
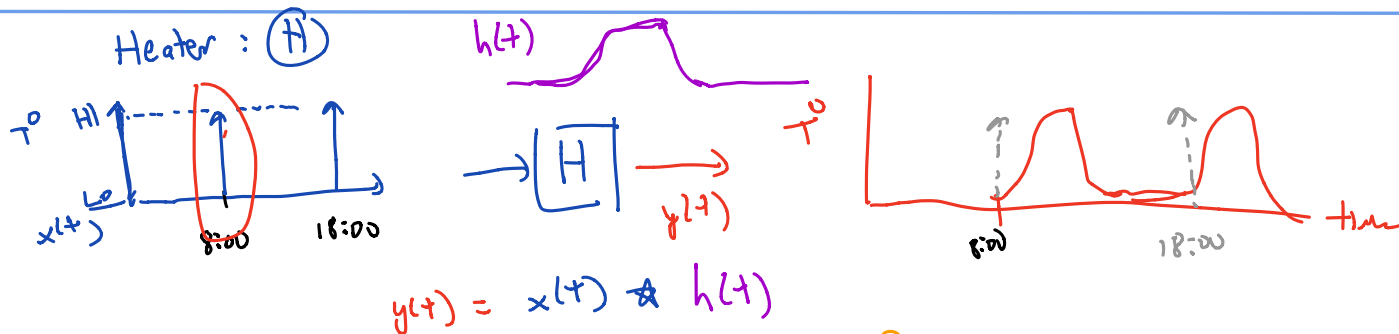
$$y(t) = x(t) * h(t)$$

Linearity  
of  $H$

Shift  
Invariance



# "Gist" of convolution - smearing



# Examples of Computing the Impulse Response

$$// y(t) = \int_{-\infty}^t x(\tau) d\tau \quad \text{LTI System} \quad \text{the Integrator}$$

$$x \rightarrow \boxed{I} \rightarrow y$$

$$y = x * h$$

What is the impulse response of this system?

- 1) Set  $x(t) = \delta(t)$
- 2) Calculate the output  $\rightarrow h(t)$

Plug-and-Chug

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

Cy: Show that (A) and (B) are equivalent.

You can also express the output in terms of the step.

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

# CYU

$$1) y(t) = x(t - 3')$$

Qst: Calculate the impulse response  
and write  $y$  in terms of  
the convolution integral

① Get IR.

$$h(t) = \delta(t - 3')$$

$$\textcircled{2} y(t) = x(t) \rightarrow h(t) = \int_{-\infty}^{\infty} x(s) h(t - s) ds = \int_{-\infty}^{\infty} x(s) \delta(t - s - 3) ds$$

Sifting Property

$$= x(t - 3')$$

# Notation of Convolution

$$\underline{y(t)} = \int_{-\infty}^{\infty} x(s) h(t-s) ds$$

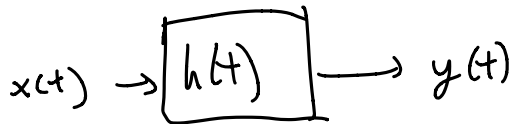
- $x$  is input
- $h$  is impulse response
- $y$  is the output

Notation

Most rigorous  $y(t) = \underline{(x \star h)(t)}$

In reality  $y = x \star h$   
 $y(t) = x(t) \star h(t)$

Block notation



$$y(t) = (x \star h)(t) = x \star h.$$

# How to Compute Convolution: flip and drag

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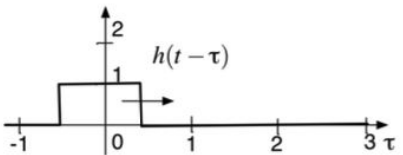
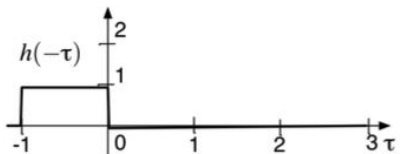
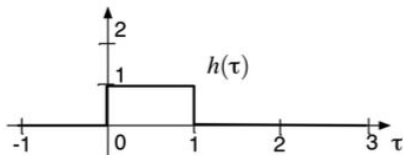
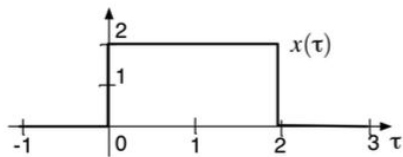
To calculate  $y(t) = (x * h)(t)$ ,

- Flip (i.e., reverse in time) the impulse response. This changes  $h(\tau)$  to  $h(-\tau)$ .
- Begin to drag the reversed time response by some amount,  $t$ . This results in  $h(t - \tau)$ .
- For a given  $t$ , multiply  $h(t - \tau)$  pointwise by  $x(\tau)$ . This produces  $x(\tau)h(t - \tau)$ .
- Integrate this product over  $\tau$ . This produces  $y(t)$  at this particular time  $t$ .

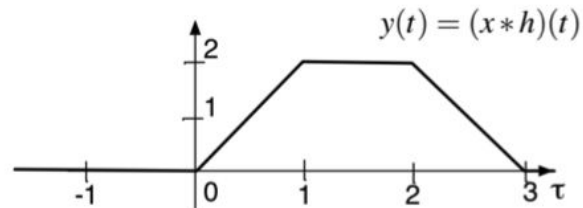
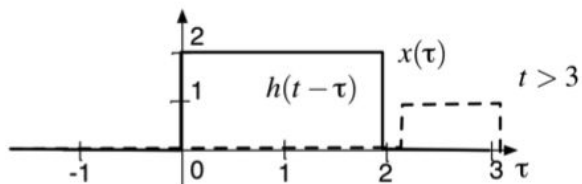
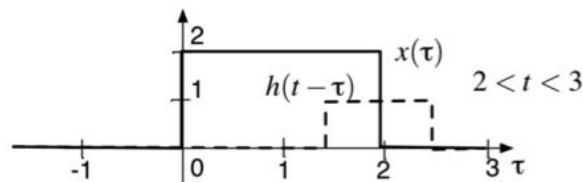
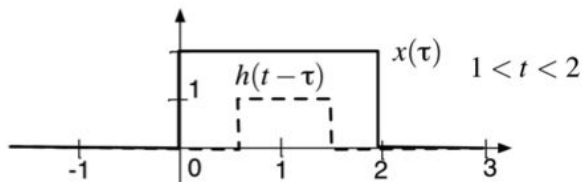
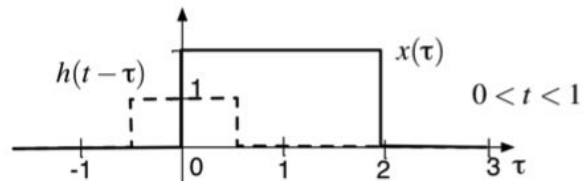
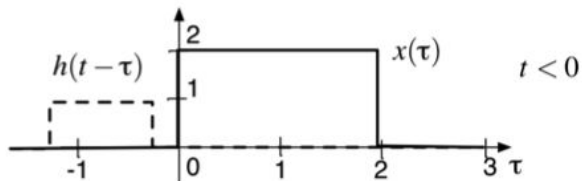
This technique is referred to as the “flip-and-drag” technique.

# How to Compute Convolution: flip and drag

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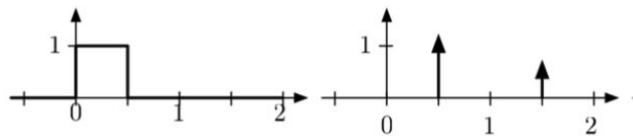
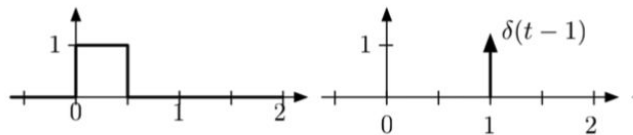
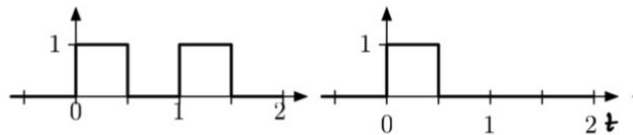
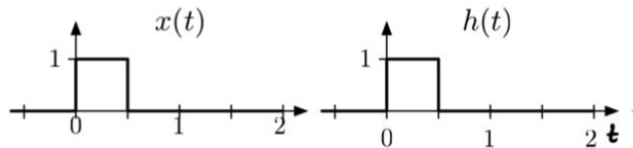


# How to Compute Convolution: flip and drag



# How to Compute Convolution: flip and drag

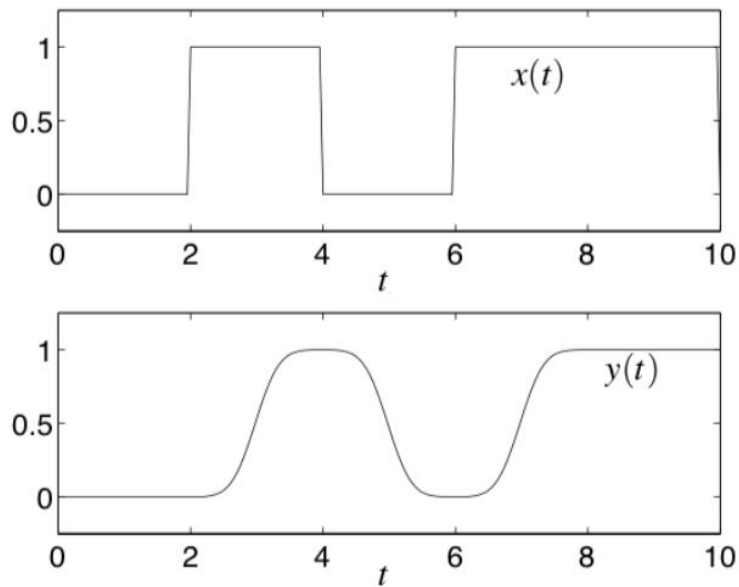
**Examples:** Try these:





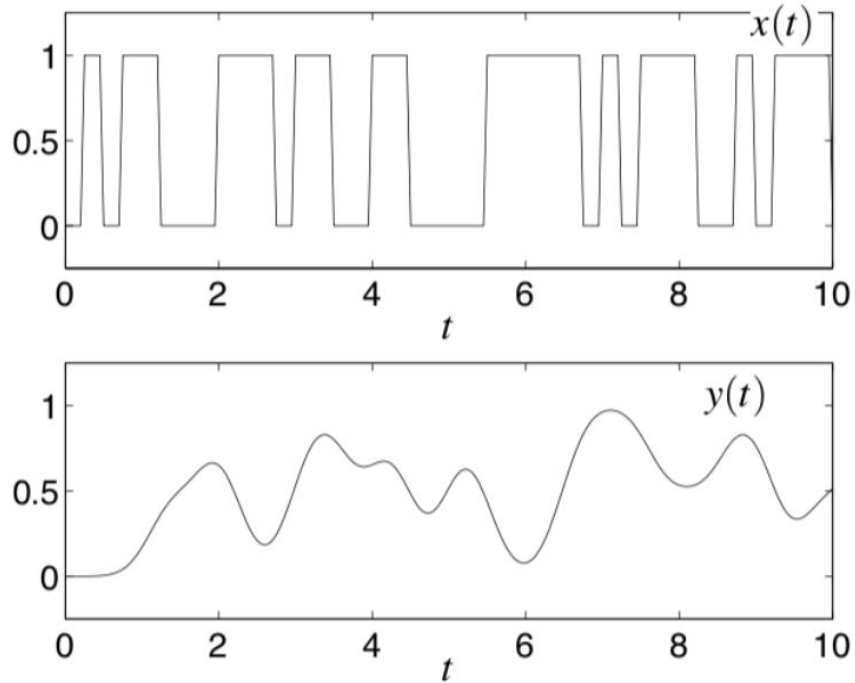
# Example: Noisy Communication

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# Example Noisy Communication

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# Causal Convolution

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## Convolution for a causal system

In a causal system,  $h(t) = 0$  for  $t < 0$ . (Why? Hint: what happens if  $h(t) \neq 0$  for some  $t < 0$ ?)

This means that  $h(t - \tau) = 0$  if  $\tau > t$ . Hence, there is no need to integrate if  $\tau$  exceeds  $t$ , since  $h(t - \tau) = 0$ . We can use this to simplify the convolution integral.

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

This equation tells us that only past and present values of  $x(\tau)$  contribute to  $y(t)$ .

# Properties of Convolution

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**Commutativity**

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

**Associativity**

$$(f * (g * h))(t) = ((f * g) * h)(t)$$

**Distributivity**

$$f * (g + h) = f * g + f * h$$

**Linearity**

$$h * (\alpha x_1 + \beta x_2) = \alpha(h * x_1) + \beta(h * x_2)$$

**Time-invariance**

# CYU Question from HW

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