## Convolution

Lecture #6 2CT.3 – 8

# Definition

Convolution is an operation on two functions of time. The following integral is the definition of convolving  $f_1(t)$  with  $f_2(t)$ :

$$g(t) = \int_{-\infty}^{\infty} f_1(\tau) f_2(t-\tau) d\tau$$

This says, first take  $f_2(t)$  and convolve it doing the following:

- 1. flip it in time :  $f_2(t) \Rightarrow f_2(-t)$
- 2. displace (shift) it in time by the amount  $\tau$  seconds:  $f_2(-t) \Rightarrow f_2(\tau t)$
- 3. take the convolved  $f_2(t)$  and multiple it by  $f_1(t)$
- 4. integrate with respect to  $\tau$  over all time which will produce another function, g(t), of time, t.

# **Properties of Convolution**

• First some shorthand:

 $\int f_1(\tau) f_2(t-\tau) d\tau \Longrightarrow f_1(t) \otimes f_2(t)$ 

• Commutative:

 $f_1(t) \otimes f_2(t) = f_2(t) \otimes f_1(t)$ 

• Associative:

 $f_1(t) \otimes [f_2(t) \otimes f_3(t)] = [f_1(t) \otimes f_2(t)] \otimes f_3(t)$ 

• Distributive:

 $f_1(t) \otimes [f_2(t) + f_3(t)] = [f_1(t) \otimes f_2(t)] + [f_1(t) \otimes f_3(t)]$ 

## A Graphical Example of How to Perform a Convolution



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## A Graphical Example of How to Perform a Convolution

We need to look at 5 cases:

1) t < 0 For this case 1, C=0 since there is no overlap.



5) t > 3T For this case 1, C=0 since there is no overlap.

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#### Matlab Code

Signals clear all; endpulse=2; 0.5 ts=.001; endpoint=10; n=-endpoint:ts:endpoint; 0 0 2 3 6 7 8 10 -1 1 4 5 9 nn=-endpoint\*2:ts:endpoint\*2; Time (Seconds) pulser=(n>=0)&(n<=endpulse);</pre> Convolution pulse=1\*pulser; tripulse= $(n \ge 0)$ & $(n \le 1)$ ; 0.4 tri=(1-n).\*tripulse; subplot(2,1,1)0.2 plot(n,tri,'r',n,pulse,'b'); title('Signals'); 0 L -1 0 2 3 4 5 6 7 8 10 1 9 xlabel('Time (Seconds)'); Shift (Seconds) axis([-1 10 min([min(tri) min(pulse)]) 1.1\*max([max(tri) max(pulse)])]); c=conv(pulse,tri)\*ts; subplot(2,1,2)plot(nn,c); title('Convolution'); xlabel('Shift (Seconds)'); axis([-1 10 min(c) 1.1\*max(c)]);

# Integration of Convolution Integral

Case 2: 0 < t < T

$$C = \int_0^t AB(1 - \frac{\tau}{T})d\tau = AB(-T)\frac{1}{2}(1 - \frac{\tau}{T})^2 \Big|_0^t = AB(-T)\frac{1}{2}\{(1 - \frac{t}{T})^2 - 1\}$$
$$= AB(-T)\frac{1}{2}\{(1 - \frac{2t}{T} + \frac{t^2}{T^2}) - 1\} = AB(-T)\frac{1}{2}\{-\frac{2t}{T} + \frac{t^2}{T^2}\}$$
$$= AB(-T)\frac{1}{2}\{-\frac{2t}{T} + \frac{t^2}{T^2}\} = AB\{t - \frac{t^2}{2T}\}$$

Case 3: T < t < 2T (From the integration from Case 2)

$$C = \int_0^T AB(1 - \frac{\tau}{T}) d\tau = AB(-T) \frac{1}{2} (1 - \frac{\tau}{T})^2 \Big|_0^T = AB\{T - \frac{T^2}{2T}\}$$
$$= AB\{T - \frac{T}{2}\} = AB\frac{T}{2}$$

Case 4: 2T < t < 3T (From the integration from Case 2)

$$C = \int_{-(2T-t)}^{T} AB(1 - \frac{\tau}{T}) d\tau = AB(-T) \frac{1}{2} (1 - \frac{\tau}{T})^2 \Big|_{-(2T-t)}^{T} = AB(-T) \frac{1}{2} \{ (1 - \frac{T}{T})^2 - (1 - \frac{-(2T-t)}{T})^2 \}$$
  
$$= AB(-T) \frac{1}{2} \{ (0)^2 - (1 + \frac{(2T-t)}{T})^2 \} = AB(T) \frac{1}{2} \{ (1 + \frac{(2T-t)}{T})^2 \}$$
  
$$= AB(T) \frac{1}{2} \{ (1 + \frac{2(2T-t)}{T} + \frac{(2T-t)^2}{T^2}) \} = AB\{ (\frac{T}{2} + (2T-t) + \frac{(2T-t)^2}{2T}) \}$$

### **Convolution and Systems**

- For an LTI system, let's define h(t) as the system response to a unit impulse source,  $\delta(t)$ .
- Then the following must be true:

 $x(t) \xrightarrow{LTI} y(t) x(t) \text{ input to the system yields output } y(t)$   $\delta(t) \xrightarrow{LTI} h(t) \ \delta(t) \text{ input to the system yields output } h(t)$   $\delta(t-k\Delta) \xrightarrow{LTI} h(t-k\Delta) \text{ time invariance, time shift by } k\Delta$   $x(k\Delta)\delta(t-k\Delta) \xrightarrow{LTI} x(k\Delta)h(t-k\Delta) \text{ scalar, multiply by } x(k\Delta)$   $\sum_{k} x(k\Delta)\delta(t-k\Delta) \xrightarrow{LTI} \sum_{k} x(k\Delta)h(t-k\Delta) \text{ superposition, add up for all value of time shift } k$   $\sum_{k} x(k\Delta)\delta(t-k\Delta) \xrightarrow{LTI} x(k\Delta)h(t-k\Delta)\Delta \text{ scalar, multiply by } \Delta$ 

#### **Convolution and Systems Continued**

Construct x(t) as the sum of k unit impulse slices. The first expression represents the kslices of the source totaled as *k* approaches infinity.

Construct y(t) as the sum of k slices of the response due to an unit impulse function:  $x(k\Delta)h(t-k\Delta)$ . The integral on the right is the convolution of x(t) and h(t).

Left Side Equals

 $\lim_{\substack{\Delta \to 0 \\ k \to \infty \\ k \Delta \to \tau}} \sum_{k} x(k\Delta) \delta(t - k\Delta) \Delta \text{ approaches } \int_{\tau} x(\tau) \delta(t - \tau) d\tau = x(t)$ Recall:  $\int_{-\infty}^{\infty} f(t)\delta(t-\tau)dt = f(\tau)$ 

**Right Side Equals** 

$$\lim_{\substack{\Delta \to 0 \\ k \to \infty \\ k \Delta \to \tau}} \sum_{k} x(k\Delta) h(t - k\Delta) \Delta \text{ approaches } \int_{\tau} x(\tau) h(t - \tau) d\tau$$

$$\sum_{k} x(k\Delta) \delta(t - k\Delta) \Delta \to \sum_{k} x(k\Delta) h(t - k\Delta) \Delta$$

$$k$$

$$k$$

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

### **Convolution and Systems Continued**

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

This result is very important since it says that if one knows the impulse response of a system then the output response for any given input source can be found by convolving the input with the impulse response.

### Impulse Response and Causality

•  $y(t_o)$  does not depend on  $x(t_i)$  that occur at times after  $t_o$ ,  $t_i > t_o$ .



one for positive  $\tau$  and one for negative  $\tau$ 

- For the positive  $\tau$  integral  $\int_{0}^{\infty} h(\tau)x(t_o \tau)d\tau$ , if  $\tau$  is positive, then  $t_o \tau < t_o$ , and for  $x(t_i = t_o - \tau) = x(t_o - \tau)$  we have  $t_i < t_o$
- For the negative  $\tau$  integral  $\int_{-\infty}^{0} h(\tau) x(t_o \tau) d\tau$ , if  $\tau$  is negative, then  $t_o \tau > t_o$ ,

and for  $x(t_i = t_o - \tau) = x(t_o - \tau)$  we have  $t_i > t_o$  THIS CAN'T HAPPEN IN A REAL CAUSAL SYSTEM.

• Therefore for y(t) to be causal this  $\int_{-\infty}^{\infty} h(\tau)x(t_o - \tau)d\tau$  must be zero and for this to happen h(t) = 0, t < 0. BME 333 Biomedical Signals and Systems - J.Schesser

### Calculating the Unit Impulse Response, h(t)



Let's first look at 2 methods:

- 1. Narrow Pulse approximation
- 2. Differentiating u(t)

Better Methods to Come

### Narrow Pulse Approximation

$$\varepsilon = \text{width of the pulse}$$

$$\frac{1}{\varepsilon} = \text{height of the pulse}$$

$$\frac{1}{\varepsilon} = \text{height of the pulse}$$
As  $\varepsilon \to 0$ , the narrow pulse  $=\frac{1}{\varepsilon} \{u(t) - u(t - \varepsilon)\} \to \delta(t)$ 
So then let's first look  
at the response to  $u(t)$ :
$$Vs u(t) = i(t)R + L \frac{di(t)}{dt}$$

$$i(t) = Vs/R (1 - e^{-(R/L)t})u(t)$$
Now we construct the narrow  
pulse response:
$$i(t) = \frac{Vs}{R} \frac{1}{\varepsilon} \{(1 - e^{-(\frac{R}{L})t})u(t) - (1 - e^{-(\frac{R}{L})(t - \varepsilon)})u(t - \varepsilon)\}$$

#### Narrow Pulse Approximation Continued

 $i(t) = (Vs/R)(1/\varepsilon) \left[ (1 - e^{-(R/L)t})u(t) - (1 - e^{-(R/L)(t-\varepsilon)})u(t-\varepsilon) \right]$ 

OR  $i(t) = \begin{cases}
0, \text{ for } t < 0 \\
(Vs/R)(1/\varepsilon) \left[(1 - e^{-(R/L)t})\right] \text{ for } 0 \le t < \varepsilon \\
(Vs/R)(1/\varepsilon) \left[(1 - e^{-(R/L)t})\right] - (1 - e^{-R/L(t-\varepsilon)})\right], \text{ for } t > \varepsilon
\end{cases}$  OR  $i(t) = \begin{cases}
0, \text{ for } t < 0 \\
(Vs/R)(1/\varepsilon) \left[(1 - e^{-(R/L)t})\right], \text{ for } 0 \le t < \varepsilon \\
(Vs/R)(1/\varepsilon) \left[(e^{R/L\varepsilon} - 1)e^{-(R/L)t})\right], \text{ for } t > \varepsilon
\end{cases}$ 

### Taylor Series Approximation for $e^x$

$$e^{ax} = \frac{(ax)^0}{0!} + \frac{(ax)^1}{1!} + \frac{(ax)^2}{2!} + \frac{(ax)^3}{3!} + \dots$$

for  $x \approx 0$ , we can drop the higher order terms:

$$e^{ax} \approx \frac{(ax)^0}{0!} + \frac{(ax)^1}{1!} = 1 + ax$$
$$e^{-\frac{R}{L}t} = 1 - \frac{R}{L}t$$
$$1 - e^{-\frac{R}{L}t} = 1 - (1 - \frac{R}{L}t) = \frac{R}{L}t$$
$$e^{\frac{R}{L}\varepsilon} - 1 = 1 + \frac{R}{L}\varepsilon - 1 = \frac{R}{L}\varepsilon$$

#### Narrow Pulse Approximation Continued

Applying the approximation for  $e^x$ , Vs=1 (the unit impulse function) and  $\alpha = R/L$  $0, \quad \text{for } t < 0$   $i(t) = \begin{cases}
(Vs/R)(1/\varepsilon) \left[(1-e^{-R/Lt})\right] = (Vs/R)(1/\varepsilon)(R/Lt) = (Vs/R)(R/Lt/\varepsilon), \quad \text{for } 0 \le t < \varepsilon \\
(Vs/R)(1/\varepsilon) \left[(e^{+R/L\varepsilon} - 1)e^{-R/Lt})\right] = (Vs/R)(1/\varepsilon)(R/L\varepsilon)e^{-R/Lt}, \quad \text{for } t > \varepsilon \\
0, \quad \text{for } t < 0 \\
(1/R)(R/Lt/\varepsilon) = R/L/R = 1/L, \quad \text{for } 0 \le t < \varepsilon \\
(1/R) R/Le^{-R/Lt}, \quad \text{for } t > \varepsilon
\end{cases}$ 

$$h(t) = \lim_{\varepsilon \to 0} i(t) = \frac{R/L}{R} e^{-R/Lt} u(t) = \frac{1}{L} e^{-R/Lt} u(t)$$

### **Differentiating the Unit Step function**

The response due to a Unit Step function is  $i(t) = Vs/R (1-e^{-\alpha t})u(t)$ and since

$$\delta(t) = \frac{du(t)}{dt}$$
, then  $h(t) = \frac{di(t)}{dt}$ 

$$i(t) = \frac{1}{R}(1 - e^{-\alpha t})u(t)$$

$$h(t) = \frac{di(t)}{dt} = \frac{1}{R}\frac{d}{dt}(1 - e^{-\alpha t})u(t) = \frac{1}{R}\frac{d}{dt}[u(t) - e^{-\alpha t}u(t)] = \frac{1}{R}[\frac{d}{dt}u(t) - \frac{d}{dt}e^{-\alpha t}u(t)]$$

$$= \frac{1}{R}[\delta(t) - \{u(t)\frac{d}{dt}e^{-\alpha t} + e^{-\alpha t}\frac{d}{dt}u(t)\}] = \frac{1}{R}[\delta(t) - \{-\alpha e^{-\alpha t}u(t) + e^{-\alpha t}\delta(t)\}]$$

$$h(t) = \frac{1}{R}[\delta(t) + \alpha e^{-\alpha t}u(t) - e^{-\alpha t}\delta(t)]$$

$$= \frac{1}{R}[(1 - e^{-\alpha t})\delta(t) + \alpha e^{-\alpha t}u(t)]$$
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# **Convolution for Discrete Systems**

- For an LTI system, let's define h[n] as the system response to a unit impulse source,  $\delta[n]$ .
- $\delta[n]=1, n=0 \text{ and } 0 \text{ for } n \neq 0$
- We have:

 $x[n] = \sum x[m]\delta[n-m]$  $y[n] = \sum x[m] h[n-m]$ 

- In addition the same convolution properties hold:
  - Commutative  $f_1[n] \otimes f_2[n] = f_2[n] \otimes f_1[n]$

  - Associative  $f_1[n] \otimes \{f_2[n] \otimes f_3[n]\} = \{f_1[n] \otimes f_2[n]\} \otimes f_3[n]$
  - Distributive  $f_1[n] \otimes \{f_2[n] + f_3[n]\} = \{f_1[n] \otimes f_2[n]\} + \{f_1[n] \otimes f_3[t]\}$

# Stability of Systems

• If a system is stable, then if the input is bounded then the output must be bounded i.e., Bounded Input, Bounded Output (BIBO), the following must be true:

 $x(t) < \infty \ \{x[n] < \infty\}, \text{ then } y(t) < \infty \ \{y[n] < \infty\}$ 

$$y(t) = \int h(\tau) x(t-\tau) d\tau < \int h(\tau) x_{\max} d\tau = x_{\max} \int h(\tau) d\tau < \infty$$

$$\{y[n] = \sum h[m]x[n-m] = x_{\max} \sum h[m] < \infty\}$$

$$OR$$

$$\int h(\tau)d\tau < \infty \quad \{\sum h[m] < \infty\}$$

- However, this is not always the case.
  - Positive Feedback causes instability

# What is needed for BIBO

- For a continuous time system, the poles of *H(p)* must lie within the left hand complex plane and not on the imaginary axis such that Re s<sub>i</sub> < 0 where s<sub>i</sub> are the poles of *H(p)*. This will assure that the free response will be damped and not grow exponentially. <u>THIS</u> <u>IS WHY WE STUDIED SOLUTIONS OF LINEAR</u> <u>ODE IN TERMS OF SOURCE-FREE AND</u> <u>SOURCE COMPONENTS.</u>
- <u>THIS IMPLIES THAT H(p) AND THE IMPULSE</u> <u>RESPONSE, h(t), MAY BE RELATED.</u>

#### What is needed for BIBO

For a discrete time system, the eigenvalues of h[n] must lie within the unit circle such that  $z_i < 1$  where  $z_i$  are the eigenvalues of  $h[n] = \sum_i A_i z_i^n$ . This will assure that free response will not diverge and  $\sum_n h[n] = \sum_n \sum_i A_i z_i^n < \infty$ . Using the formula for the partial sums of a geometric series, where N is the number of roots of the Characteristic Equation  $\lim_{n \to \infty} \sum_{i=1}^{L-1} h[n] = \lim_{n \to \infty} \sum_{i=1}^{N} \sum_{i=1}^{L-1} z_i^n = \lim_{n \to \infty} \sum_{i=1}^{N} \frac{1-z_i^L}{1-z_i} \rightarrow \sum_{i=1}^{N} \frac{1}{1-z_i}$ ; provided  $|z_i| < 1$ 

$$\begin{split} \lim_{L \to \infty} \sum_{n=0}^{L-1} h[n] &= \lim_{L \to \infty} \sum_{n=0}^{L-1} \sum_{i=1}^{N} z_i^{\ n} = \lim_{L \to \infty} \sum_{i=1}^{N} \sum_{n=0}^{L-1} z_i^{\ n} = \lim_{L \to \infty} \sum_{i=1}^{N} \frac{1-z_i^{\ L}}{1-z_i} \to \sum_{i=1}^{N} \frac{1}{1-z_i}; \text{ provided } |z_i| < 1 \\ \text{if } |z_i| &\geq 1, \lim_{L \to \infty} \sum_{n=0}^{L-1} \sum_{i=1}^{N} z_i^{\ n} = \lim_{L \to \infty} \sum_{i=1}^{N} \left| \frac{1-z_i^{\ L}}{1-z_i} \right| \to \infty \\ z_i &= |z_i| \angle \text{angle}(z_i) \\ z_i^{\ L} &= |z_i|^L \angle (L \times \text{angle}(z_i)) \\ \lim_{L \to \infty} z_i^{\ L} &= \lim_{L \to \infty} \{ |z_i|^L \angle (L \times \text{angle}(z_i)) \} = \lim_{L \to \infty} \{ |z_i|^L \} \angle \{ -\pi < L \times \text{angle}(z_i) \le \pi \} \\ \lim_{L \to \infty} \{ |z_i|^L \} &= 0; z_i < 1 \text{ and } \lim_{L \to \infty} \{ |z_i|^L \} \to \infty; z_i \ge 1 \end{split}$$

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## Homework

- Convolution Verify your all your results of these convolution problems using Matlab and its conv function.
  - Problem (1)
    - Assume that a system response is given by the following:

• Sketch the response to a) u(t), b) u(t)-u(t-a) for a=0.5, a=1, and a=5, and c) evaluate  $e^{-t} u(t)$  at t=1 and t=2

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- Problem (2)
  - Assume that a system response is given by the following:



• Evaluate the response to  $te^{-t} u(t)$  at t=1 and t=3

## Homework

- Stability
  - Determine the stability of the following systems with poles in the complex plane, describe the form of the transient response:



• 2CT.3.1, 2CT.3.2