

Signals and Systems UNIT 2

Ripal Patel

Assistant Professor,
Dr.Ambedkar Institute of Technology, Bangalore.

ripal.patel@dr-ait.org

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A continuous-time (discrete-time) system H is an operator that transfer the input $x(t)$ ($x[n]$) into the output $y(t)$ ($y[n]$). We denote the process by

$$x(t) \implies \left[H \right] \implies y(t)$$

$$x(n) \implies \left[H \right] \implies y(n)$$

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

Example: $y(t) = x(t) + 1$

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

Example: $y(n) = x(n)^2$

- A system H is memoryless if the value $y(t_0)$ (i.e., $y(t = t_0)$) only depends on the value $x(t_0)$ for any t_0 .

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- Current time t :
Current input: $x(t)$
Past input: $x(t-1), x(t-2), \dots, x(t-k)$
Future input: $x(t+1), x(t+2), \dots, x(t+k)$

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- Example: $y(t) = x(t-1)$ is a system with memory since $y(t_0) = x(t_0-1)$, e.g., $y(0) = x(-1)$. $y(t_0)$ depends on $x(t)$ at $t = t_0-1$, not at t_0 .

- A system H is memoryless if the value $y(t_0)$ (i.e., $y(t = t_0)$) only depends on the value $x(t_0)$ for any t_0 .
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- Example: $y(t) = x^2(t)$ is memoryless since $y(t_0) = x^2(t_0)$ for t_0 .
- Example: $y(t) = x(t-1)$ is a system with memory since $y(t_0) = x(t_0-1)$, e.g., $y(0) = x(-1)$. $y(t_0)$ depends on $x(t)$ at $t = t_0-1$, not at t_0 .
- In other words, output $y(t)$ at current time $t = t_0$ is only affected by input $x(t)$ at current time $t = t_0$

- Problem: $y(n) = nx(n)$

- Problem: $y(n) = nx(n)$
- Answer: Memoryless

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- Problem: $y(n) = x(n)x(n-1)$

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- Answer: Memory
- Problem: $y(t) = x(2-t)$

- Problem: $y(n) = nx(n)$
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- Problem: $y(t) = \int_{-\infty}^{t/2} x(\tau) d\tau$

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- Problem: $y(t) = \int_{-\infty}^{t/2} x(\tau) d\tau$
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- Problem: $y(t) = \frac{d}{dt}x(t)$

- Problem: $y(n) = nx(n)$
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- Problem: $y(t) = \frac{d}{dt}x(t)$
- Answer: Memory

- A system H is causal if the value $y(t_0)$ only depends on $x(t) : t \leq t_0$.
- I.e., current output $y(t)$ is produced by current input $x(t)$ and past input $x(t-1), x(t-2), \dots, x(t-k)$, not future input $x(t+1), x(t+2), \dots, x(t+k)$.
- The system $y[n] = x[n-1]$ is causal ($y[0] = x[-1]$)
- The system $y[n] = x[n+1]$ is non-causal ($y[0] = x[1]$)
- The system $y(t) = x(t+a)$ is causal if $a \leq 0$ and is non-causal if $a > 0$

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System
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System Classification

Memory vs.
Memoryless**Causal vs.
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Non-stable

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Convolution Sum

- Problem: $y(n) = nx(n)$

- Problem: $y(n) = nx(n)$
- Answer: Causal

- Problem: $y(n) = nx(n)$
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- Problem: $y(n) = x(n)x(n - 1)$

- Problem: $y(n) = nx(n)$
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- Problem: $y(n) = x(n)x(n-1)$
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- Problem: $y(t) = x(2-t)$

- Problem: $y(n) = nx(n)$
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- Problem: $y(n) = x(n)x(n-1)$
- Answer: Causal
- Problem: $y(t) = x(2-t)$
- Answer: Non-causal

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- Problem: $y(t) = x(2-t)$
- Answer: Non-causal
- Problem: $y(t) = \int_{-\infty}^{t/2} x(\tau) d\tau$
- Answer: Causal
- Problem: $y(t) = \frac{d}{dt}x(t)$

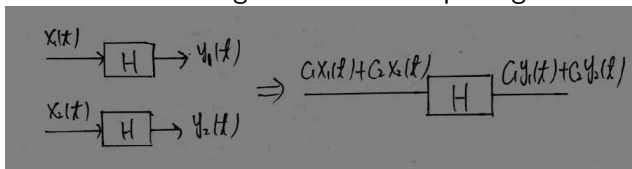
- Problem: $y(n) = nx(n)$
- Answer: Causal
- Problem: $y(n) = x(n)x(n - 1)$
- Answer: Causal
- Problem: $y(t) = x(2 - t)$
- Answer: Non-causal
- Problem: $y(t) = \int_{-\infty}^{t/2} x(\tau) d\tau$
- Answer: Causal
- Problem: $y(t) = \frac{d}{dt}x(t)$
- Answer: Causal

- H is called linear if H has the superposition property:

$$H[a_1x_1(t) + a_2x_2(t)] = a_1H[x_1(t)] + a_2H[x_2(t)]$$

where, $y_1(t) = H[x_1(t)]$ and $y_2(t) = H[x_2(t)]$

- The response of a weighted sum of input signals is equal to the same as weighted sum of output signals.



$$y(n) = x(n - 3)$$

- Two output signals

$$y_1(n) = H[x_1(n)] = x_1(n - 3)$$

$$y_2(n) = H[x_2(n)] = x_2(n - 3)$$

-

$$H[a_1x_1(n) + a_2x_2(n)] = a_1x_1(n - 3) + a_2x_2(n - 3)$$

$$= a_1H[x_1(n)] + a_2H[x_2(n)]$$

$$a_1y_1(n) + a_2y_2(n)$$

Hence, system is LINEAR.

$$y(t) = x^2(t)$$

- Two output signals

$$y_1(n) = H[x_1(t)] = x_1^2(t)$$

$$y_2(n) = H[x_2(t)] = x_2^2(t)$$

- LHS:

$$H[a_1x_1(t) + a_2x_2(t)] = [a_1x_1(t) + a_2x_2(t)]^2$$

- RHS:

$$a_1H[x_1(t)] + a_2H[x_2(t)] = a_1x_1^2(t) + a_2x_2^2(t)$$

LHS \neq RHS. Hence, system is NON-LINEAR.

$$y(n) = nx(n)$$

- Two output signals

$$y_1(n) = H[x_1(n)] = nx_1(n)$$

$$y_2(n) = H[x_2(n)] = nx_2(n)$$

-

$$H[a_1x_1(n) + a_2x_2(n)] = a_1nx_1(n) + a_2nx_2(n)$$

$$= a_1H[x_1(n)] + a_2H[x_2(n)]$$

$$a_1y_1(n) + a_2y_2(n)$$

Hence, system is LINEAR.

$$y(n) = x(n)x(n-1)$$

- Two output signals

$$y_1(n) = H[x_1(n)] = x_1(n)x_1(n-1)$$

$$y_2(n) = H[x_2(n)] = x_2(n)x_2(n-1)$$

-

$$H[a_1x_1(n) + a_2x_2(n)] = a_1x_1(n)x_1(n-1) + a_2x_2(n)x_2(n-1)$$

$$= a_1H[x_1(n)] + a_2H[x_2(n)]$$

$$a_1y_1(n) + a_2y_2(n)$$

Hence, system is LINEAR.

$$y(t) = \frac{d}{dt}x(t)$$

- Two output signals

$$y_1(n) = H[x_1(t)] = \frac{d}{dt}x_1(t)$$

$$y_2(n) = H[x_2(t)] = \frac{d}{dt}x_2(t)$$

- LHS:

$$\begin{aligned} H[a_1x_1(t) + a_2x_2(t)] &= \frac{d}{dt}[a_1x_1(t) + a_2x_2(t)] \\ &= a_1\frac{d}{dt}x_1(t) + a_2\frac{d}{dt}x_2(t) \end{aligned}$$

- RHS:

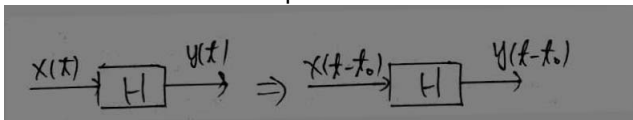
$$a_1H[x_1(t)] + a_2H[x_2(t)] = a_1\frac{d}{dt}x_1(t) + a_2\frac{d}{dt}x_2(t)$$

LHS = RHS. Hence, system is LINEAR.

- H is called time-invariant if the following is true:

$$H[x(t)] = y(t) \implies H[x(t - t_0)] = y(t - t_0)$$

- I.e., a time-shift t_0 in the input $x(t)$ results in an identical time-shift t_0 in the output.



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$$y(t) = e^{x(t)}$$



$$y(t) = e^{x(t)}$$



$$H(x(t - t_0) = e^{x(t-t_0)})$$

$$y(t - t_0) = e^{x(t-t_0)}$$



$$y(t) = e^{x(t)}$$



$$H(x(t - t_0)) = e^{x(t-t_0)}$$

$$y(t - t_0) = e^{x(t-t_0)}$$

- Time-invariant

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$$y(t) = g(t)x(t)$$

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$$y(t) = g(t)x(t)$$

$$H(x(t - t_0) = g(t)x(t - t_0))$$

$$y(t - t_0) = g(t - t_0)x(t - t_0)$$

-

$$y(t) = g(t)x(t)$$

-

$$H(x(t - t_0) = g(t)x(t - t_0))$$

$$y(t - t_0) = g(t - t_0)x(t - t_0)$$

- Time-variant

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$$y(n) = x(n - k)$$



$$y(n) = x(n - k)$$



$$H(x(n - k - n_0)) = x(n - k - n_0)$$

$$y(n - k - n_0) = x(n - k - n_0)$$

-

$$y(n) = x(n - k)$$

-

$$H(x(n - k - n_0)) = x(n - k - n_0)$$

$$y(n - k - n_0) = x(n - k - n_0)$$

- Time-Invariant

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$$y(t) = x(2t)$$

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$$y(t) = x(2t)$$

$$H(x(t - t_0) = x(2t - t_0))$$

$$y(t - t_0) = x(2(t - t_0))$$



$$y(t) = x(2t)$$



$$H(x(t - t_0) = x(2t - t_0))$$

$$y(t - t_0) = x(2(t - t_0))$$

- Time-variant

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$$y(t) = \sin[x(t)]$$



$$y(t) = \sin[x(t)]$$



$$H(x(t - t_0) = \sin[x(t - t_0)]$$

$$y(t - t_0) = \sin[x(t - t_0)]$$

-

$$y(t) = \sin[x(t)]$$

-

$$H(x(t - t_0) = \sin[x(t - t_0)]$$

$$y(t - t_0) = \sin[x(t - t_0)]$$

- Time-invariant

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$$y(t) = x[\cos t]$$

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$$y(t) = x[\cos t]$$

$$H(x(t - t_0) = x[\cos t - t_0]$$

$$y(t - t_0) = x[\cos(t - t_0)]$$



$$y(t) = x[\cos t]$$



$$H(x(t - t_0) = x[\cos t - t_0]$$

$$y(t - t_0) = x[\cos(t - t_0)]$$

- Time-variant

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$$y(t) = \cos t \cdot x(t)$$



$$y(t) = \cos t.x(t)$$



$$H(x(t - t_0) = \cos t.x(t - t_0))$$

$$y(t - t_0) = \cos(t - t_0).x(t - t_0)$$

-

$$y(t) = \cos t.x(t)$$

-

$$H(x(t - t_0) = \cos t.x(t - t_0))$$

$$y(t - t_0) = \cos(t - t_0).x(t - t_0)$$

- Time-variant

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- Problem: $y(t) = \int_{-\infty}^{t/2} x(\tau) d\tau$

- Problem: $y(t) = \int_{-\infty}^{t/2} x(\tau) d\tau$

$$\begin{aligned}
 y_d(t) &= \int_{-\infty}^{t/2} x(\tau - d) d\tau \\
 &= \int_{-\infty}^{t/2-d} x(s) ds \\
 &= \int_{-\infty}^{(t-2d)/2} x(s) ds \\
 &= y(t - 2d).
 \end{aligned}$$

-

- Problem: $y(t) = \int_{-\infty}^{t/2} x(\tau) d\tau$

$$\begin{aligned}
 y_d(t) &= \int_{-\infty}^{t/2} x(\tau - d) d\tau \\
 &= \int_{-\infty}^{t/2-d} x(s) ds \\
 &= \int_{-\infty}^{(t-2d)/2} x(s) ds \\
 &= y(t - 2d).
 \end{aligned}$$

-
- Time-variant: Therefore, it does not obey the time-invariance condition.

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

Time-invariant vs. Time-variant EXAMPLE

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

The integrator system is also a time-invariant system. To prove this, we replace $x(\tau)$ in (9.30) by $x(\tau - t_0)$ obtaining the output $w(t)$

$$w(t) = \int_{-\infty}^t x(\tau - t_0) d\tau \quad (9.35)$$

Now, to prove time-invariance, we must manipulate the integral in (9.35) into a form that is recognizable in terms of the original output $y(t)$. This is done by changing the “dummy variable” of integration to $\sigma = \tau - t_0$. In this substitution, $d\tau$ is replaced by $d\sigma$, the lower limit $\tau = -\infty$ becomes $\sigma = -\infty$, and the upper limit $\tau = t$ becomes $\sigma = t - t_0$. Therefore (9.35) becomes

$$w(t) = \int_{-\infty}^{t-t_0} x(\sigma) d\sigma$$

- and it is now clear that $w(t) = y(t - t_0)$, so the integrator system is seen to be time-invariant. ■

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Convolution Sum

- Problem: $y(t) = \frac{d}{dt}x(t)$

- Problem: $y(t) = \frac{d}{dt}x(t)$
- The derivative of a time-shifted signal is

$$y_d(t) = \frac{d}{dt}[x(t-d)] = \frac{dx}{dt}(t-d) \frac{d}{dt}(t-d) = \frac{dx}{dt}(t-d) = y(t-d).$$

- Problem: $y(t) = \frac{d}{dt}x(t)$
- The derivative of a time-shifted signal is

$$y_d(t) = \frac{d}{dt}[x(t-d)] = \frac{dx}{dt}(t-d) \frac{d}{dt}(t-d) = \frac{dx}{dt}(t-d) = y(t-d).$$

- Time-invariant

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Convolution Sum

- Problem: $y(n) = 2x(n)u(n)$

- Problem: $y(n) = 2x(n)u(n)$
- Not Time Invariant. The time-invariance condition does not hold, because the signal that is being multiplied by $x(n)$ varies with time.

- Problem: $y(t) = x(2 - t)$

- Problem: $y(t) = x(2 - t)$

-

$$H(x(t - t_0) = x(2 - t - t_0))$$

$$y(t - t_0) = x(2 - (t - t_0)) = x(2 - t + t_0)$$

- Problem: $y(t) = x(2 - t)$

-

$$H(x(t - t_0) = x(2 - t - t_0))$$

$$y(t - t_0) = x(2 - (t - t_0)) = x(2 - t + t_0)$$

- Time-variant

- Problem: $y(n) = nx(n)$

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Impulse Response
Convolution Sum

- Problem: $y(n) = nx(n)$
- Answer: Time-variant

- Problem: $y(n) = nx(n)$
- Answer: Time-variant
- Problem: $y(n) = x(n)x(n - 1)$

- Problem: $y(n) = nx(n)$
- Answer: Time-variant
- Problem: $y(n) = x(n)x(n - 1)$
- Answer: Time-invariant

- Problem: $y(n) = nx(n)$
- Answer: Time-variant
- Problem: $y(n) = x(n)x(n - 1)$
- Answer: Time-invariant
- Problem: $y(t) = x(2 - t)$

- Problem: $y(n) = nx(n)$
- Answer: Time-variant
- Problem: $y(n) = x(n)x(n - 1)$
- Answer: Time-invariant
- Problem: $y(t) = x(2 - t)$
- Answer: Time-variant

- A system is said to be Bounded Input Bounded Output (BIBO) stable if and only if every bounded input results in bounded output.
- System H is BIBO stable if $|x(t)| \leq M_x < \infty$ then $|y(t)| \leq M_y < \infty$ for all t

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- Problem: $y(t) = x(t - t_0)$

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- Problem: $y(t) = x(t - t_0)$
- Bounded Input

$$|x(t)| \leq M_x < \infty$$

- Problem: $y(t) = x(t - t_0)$

- Bounded Input

$$|x(t)| \leq M_x < \infty$$

- By shifting input $x(t)$ by t_0

$$|x(t - t_0)| \leq M_x < \infty$$

- Problem: $y(t) = x(t - t_0)$

- Bounded Input

$$|x(t)| \leq M_x < \infty$$

- By shifting input $x(t)$ by t_0

$$|x(t - t_0)| \leq M_x < \infty$$

- Taking magnitude on both sides of input-output relationship,

$$|y(t)| = |x(t - t_0)| \leq M_x < \infty$$

- Problem: $y(t) = x(t - t_0)$

- Bounded Input

$$|x(t)| \leq M_x < \infty$$

- By shifting input $x(t)$ by t_0

$$|x(t - t_0)| \leq M_x < \infty$$

- Taking magnitude on both sides of input-output relationship,

$$|y(t)| = |x(t - t_0)| \leq M_x < \infty$$

- BIBO condition satisfied. Hence, stable.

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- Problem: $y(n) = x(-n)$

- Problem: $y(n) = x(-n)$
- Bounded Input

$$|x(n)| \leq M_x < \infty$$

- Problem: $y(n) = x(-n)$

- Bounded Input

$$|x(n)| \leq M_x < \infty$$

- By time reversing input $x(n)$ by

$$|x(-n)| \leq M_x < \infty$$

- Problem: $y(n) = x(-n)$

- Bounded Input

$$|x(n)| \leq M_x < \infty$$

- By time reversing input $x(n)$ by

$$|x(-n)| \leq M_x < \infty$$

- Taking magnitude on both sides of input-output relationship,

$$|y(n)| = |x(-n)| \leq M_x < \infty$$

- Problem: $y(n) = x(-n)$

- Bounded Input

$$|x(n)| \leq M_x < \infty$$

- By time reversing input $x(n)$ by

$$|x(-n)| \leq M_x < \infty$$

- Taking magnitude on both sides of input-output relationship,

$$|y(n)| = |x(-n)| \leq M_x < \infty$$

- BIBO condition satisfied. Hence, stable.

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Impulse Response
Convolution Sum

- Problem: $y(t) = [\sin 6t]x(t)$

- Problem: $y(t) = [\sin 6t]x(t)$
- Bounded Input

$$|x(t)| \leq M_x < \infty$$

- Problem: $y(t) = [\sin 6t]x(t)$

- Bounded Input

$$|x(t)| \leq M_x < \infty$$

- Taking magnitude on both sides of input-output relationship,

$$|y(t)| = |\sin 6t||x(t)|$$

Now,

$$|\sin 6t| \leq 1$$

$$|x(t)| \leq M_x < \infty$$

$y(t)$ is also bounded.

- Problem: $y(t) = [\sin 6t]x(t)$

- Bounded Input

$$|x(t)| \leq M_x < \infty$$

- Taking magnitude on both sides of input-output relationship,

$$|y(t)| = |\sin 6t||x(t)|$$

Now,

$$|\sin 6t| \leq 1$$

$$|x(t)| \leq M_x < \infty$$

$y(t)$ is also bounded.

- BIBO condition satisfied. Hence, stable.

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- Problem: $y(n) = x(n) + n$

- Problem: $y(n) = x(n) + n$
- Bounded Input

$$|x(n)| \leq M_x < \infty$$

- Problem: $y(n) = x(n) + n$

- Bounded Input

$$|x(n)| \leq M_x < \infty$$

- Taking magnitude on both sides of input-output relationship,

$$|y(n)| = |x(n) + n| = |x(n)| + |n|$$

As $n \rightarrow \infty$, $y(n) \rightarrow \infty$ means output is not bounded.

- Problem: $y(n) = x(n) + n$

- Bounded Input

$$|x(n)| \leq M_x < \infty$$

- Taking magnitude on both sides of input-output relationship,

$$|y(n)| = |x(n) + n| = |x(n)| + |n|$$

As $n \rightarrow \infty$, $y(n) \rightarrow \infty$ means output is not bounded.

- BIBO condition not-satisfied. Hence, Unstable.

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- Problem: $y(t) = x\left(\frac{t}{2}\right)$

- Problem: $y(t) = x(\frac{t}{2})$
- Bounded Input

$$|x(t)| \leq M_x < \infty$$

- Problem: $y(t) = x(\frac{t}{2})$

- Bounded Input

$$|x(t)| \leq M_x < \infty$$

- By shifting input $x(t)$ by t_0

$$|x(\frac{t}{2})| \leq M_x < \infty$$

- Problem: $y(t) = x\left(\frac{t}{2}\right)$

- Bounded Input

$$|x(t)| \leq M_x < \infty$$

- By shifting input $x(t)$ by t_0

$$\left|x\left(\frac{t}{2}\right)\right| \leq M_x < \infty$$

- Taking magnitude on both sides of input-output relationship,

$$|y(t)| = \left|x\left(\frac{t}{2}\right)\right| \leq M_x < \infty$$

- Problem: $y(t) = x\left(\frac{t}{2}\right)$

- Bounded Input

$$|x(t)| \leq M_x < \infty$$

- By shifting input $x(t)$ by t_0

$$\left|x\left(\frac{t}{2}\right)\right| \leq M_x < \infty$$

- Taking magnitude on both sides of input-output relationship,

$$|y(t)| = \left|x\left(\frac{t}{2}\right)\right| \leq M_x < \infty$$

- BIBO condition satisfied. Hence, stable.

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- Problem: $y(t) = e^{x(t)}$

- Problem: $y(t) = e^{x(t)}$
- Bounded Input

$$|x(t)| \leq M_x < \infty$$

- Problem: $y(t) = e^{x(t)}$

- Bounded Input

$$|x(t)| \leq M_x < \infty$$

- By shifting input $x(t)$ by t_0

$$|e^{x(t)}| \leq M_x < \infty$$

- Problem: $y(t) = e^{x(t)}$

- Bounded Input

$$|x(t)| \leq M_x < \infty$$

- By shifting input $x(t)$ by t_0

$$|e^{x(t)}| \leq M_x < \infty$$

- Taking magnitude on both sides of input-output relationship,

$$|y(t)| = |e^{x(t)}| \leq M_x < \infty$$

- Problem: $y(t) = e^{x(t)}$

- Bounded Input

$$|x(t)| \leq M_x < \infty$$

- By shifting input $x(t)$ by t_0

$$|e^{x(t)}| \leq M_x < \infty$$

- Taking magnitude on both sides of input-output relationship,

$$|y(t)| = |e^{x(t)}| \leq M_x < \infty$$

- BIBO condition satisfied. Hence, stable.

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- Problem: $y(n) = \cos[x(n)]$

- Problem: $y(n) = \cos[x(n)]$
- Bounded Input

$$|x(n)| \leq M_x < \infty$$

- Problem: $y(n) = \cos[x(n)]$

- Bounded Input

$$|x(n)| \leq M_x < \infty$$

- Taking magnitude on both sides of input-output relationship,

$$|y(n)| = |\cos[x(n)]|$$

As $|\cos[x(n)]| \leq 1$, $y(n) \rightarrow \infty$ means output is bounded.

- Problem: $y(n) = \cos[x(n)]$

- Bounded Input

$$|x(n)| \leq M_x < \infty$$

- Taking magnitude on both sides of input-output relationship,

$$|y(n)| = |\cos[x(n)]|$$

As $|\cos[x(n)]| \leq 1$, $y(n) \rightarrow \infty$ means output is bounded.

- BIBO condition satisfied. Hence, stable.

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- Problem: $y(n) = x(n)^2$

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- Problem: $y(n) = x(n)^2$
- Bounded Input

$$|x(n)| \leq M_x < \infty$$

- Problem: $y(n) = x(n)^2$

- Bounded Input

$$|x(n)| \leq M_x < \infty$$

- Taking magnitude on both sides of input-output relationship,

$$|y(n)| = |x(n)^2|$$

As $|x(n)^2| \leq M_x^2 \leq \infty$ means output is bounded.

- Problem: $y(n) = x(n)^2$
- Bounded Input

$$|x(n)| \leq M_x < \infty$$

- Taking magnitude on both sides of input-output relationship,

$$|y(n)| = |x(n)^2|$$

As $|x(n)^2| \leq M_x^2 \leq \infty$ means output is bounded.

- BIBO condition satisfied. Hence, stable.

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Convolution Sum

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

- Problem: $y(t) = \int_{-\infty}^t x(\tau) d\tau$
- Bounded Input

$$|x(t)| \leq M_x < \infty$$

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

- Bounded Input

$$|x(t)| \leq M_x < \infty$$

- Taking magnitude on both sides of input-output relationship,

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

let's take $x(t) = u(t)$

$$|u(t)| = 1$$

$$|y(t)| = \left| \int_{-\infty}^t x(\tau) d\tau \right| = \left| \int_{-\infty}^t 1 d\tau \right| = [\tau]_{-\infty}^t = t + \infty$$

if $t \rightarrow \infty$ then $|y(t)| \rightarrow \infty$

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

- Bounded Input

$$|x(t)| \leq M_x < \infty$$

- Taking magnitude on both sides of input-output relationship,

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

let's take $x(t) = u(t)$

$$|u(t)| = 1$$

$$|y(t)| = \left| \int_{-\infty}^t x(\tau) d\tau \right| = \left| \int_{-\infty}^t 1 d\tau \right| = [\tau]_{-\infty}^t = t + \infty$$

if $t \rightarrow \infty$ then $|y(t)| \rightarrow \infty$

- BIBO condition not satisfied. Hence, Unstable.

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- A system is said to be invertible if the input to the system may be uniquely determined from the output.

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Impulse Response
Convolution Sum

- A system is said to be invertible if the input to the system may be uniquely determined from the output.
- Distinct inputs produce distinct outputs

- A system is said to be invertible if the input to the system may be uniquely determined from the output.
- Distinct inputs produce distinct outputs
- A system is said to be invertible, if the inverse of that system exists.

$$y(t) = T[x(t)] \Rightarrow x(t) = T_i[y(t)]$$

$$y(t) = T[x(t)] = 10x(t) \Rightarrow x(t) = T_i[y(t)] = 0.1y(t)$$

To test for invertibility, we use two different techniques:

- We may show that a system is invertible by designing an inverse system that uniquely recovers the input from output.

To test for invertibility, we use two different techniques:

- We may show that a system is invertible by designing an inverse system that uniquely recovers the input from output.
- We may show that system is not invertible by finding two different inputs that produce the same output.

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Convolution Sum



$$y[n] = 2x[n]$$

-

$$y[n] = 2x[n]$$

- For given output $y[n]$, we may recover the input using

$$x[n] = \frac{1}{2}y[n]$$

-

$$y[n] = 2x[n]$$

- For given output $y[n]$, we may recover the input using

$$x[n] = \frac{1}{2}y[n]$$

- Hence, system is invertible.

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$$y(n) = \text{Re}[x(n)]$$

-

$$y(n) = \text{Re}[x(n)]$$

- Two different inputs can produce same output.

$$y(n) = \text{Re}[x(n)] = \text{Re}[2 - j2] = 2$$

$$y(n) = \text{Re}[x(n)] = \text{Re}[2 + j2] = 2$$

-

$$y(n) = \text{Re}[x(n)]$$

- Two different inputs can produce same output.

$$y(n) = \text{Re}[x(n)] = \text{Re}[2 - j2] = 2$$

$$y(n) = \text{Re}[x(n)] = \text{Re}[2 + j2] = 2$$

- Hence, system is not-invertible.

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Convolution Sum



$$y(t) = \cos[x(t)]$$

- $y(t) = \cos[x(t)]$
- Two different inputs can produce same output.

$$y(t) = \cos[x(t)] = \cos[0] = 1$$

$$y(t) = \cos[x(t)] = \cos[2\pi] = 1$$

-

$$y(t) = \cos[x(t)]$$

- Two different inputs can produce same output.

$$y(t) = \cos[x(t)] = \cos[0] = 1$$

$$y(t) = \cos[x(t)] = \cos[2\pi] = 1$$

- Hence, system is not-invertible.

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$$y(n) = x[n] + 0.5x[n - 1]$$

- $$y(n) = x[n] + 0.5x[n - 1]$$
- For the given system the inverse of the system is possible,

$$x(n) = y[n] - 0.5x[n - 1]$$

- $$y(n) = x[n] + 0.5x[n - 1]$$
- For the given system the inverse of the system is possible,

$$x(n) = y[n] - 0.5x[n - 1]$$

- Hence, system is Invertible.

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$$y(n) = x^2[n]$$

-

$$y(n) = x^2[n]$$

- Two different inputs can produce same output.

$$y(n) = x^2[n] = (-2)^2 = 4$$

$$y(n) = x^2[n] = (2)^2 = 4$$

-

$$y(n) = x^2[n]$$

- Two different inputs can produce same output.

$$y(n) = x^2[n] = (-2)^2 = 4$$

$$y(n) = x^2[n] = (2)^2 = 4$$

- Hence, system is not-invertible.

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Impulse Response
Convolution Sum



$$y(t) = 2^{-x(t)}$$

- $$y(t) = 2^{-x(t)}$$
- For the given system, input can be find out using

$$x(t) = \log_2 y(t)$$

- $$y(t) = 2^{-x(t)}$$
- For the given system, input can be find out using

$$x(t) = \log_2 y(t)$$

- Hence, system is not-invertible.

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Convolution Sum



$$y[n] = x[n] - x[n - 1]$$

-

$$y[n] = x[n] - x[n - 1]$$

- Two different inputs can produce same output.

$$y[n] = x[n] - x[n - 1] = 5 - 3 = 2$$

$$y[n] = x[n] - x[n - 1] = 9 - 7 = 2$$

- $$y[n] = x[n] - x[n - 1]$$
- Two different inputs can produce same output.

$$y[n] = x[n] - x[n - 1] = 5 - 3 = 2$$

$$y[n] = x[n] - x[n - 1] = 9 - 7 = 2$$

- Hence, system is not-invertible.

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- $$y(t) = x^4(t)$$

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Convolution Sum

-

$$y(t) = x^4(t)$$

- Not-invertible



$$y(t) = x^4(t)$$

- Not-invertible



$$y(t) = Ax(t)$$

-

$$y(t) = x^4(t)$$

- Not-invertible

-

$$y(t) = Ax(t)$$

- Invertible

-

$$y(t) = x^4(t)$$

- Not-invertible

-

$$y(t) = Ax(t)$$

- Invertible

-

$$y(t) = Ax(t) + B$$

-

$$y(t) = x^4(t)$$

- Not-invertible

-

$$y(t) = Ax(t)$$

- Invertible

-

$$y(t) = Ax(t) + B$$

- Invertible

-

$$y(t) = x^4(t)$$

- Not-invertible

-

$$y(t) = Ax(t)$$

- Invertible

-

$$y(t) = Ax(t) + B$$

- Invertible

-

$$y(n) = x(-n)$$

-

$$y(t) = x^4(t)$$

- Not-invertible

-

$$y(t) = Ax(t)$$

- Invertible

-

$$y(t) = Ax(t) + B$$

- Invertible

-

$$y(n) = x(-n)$$

- Invertible

-

$$y(t) = x^4(t)$$

- Not-invertible

-

$$y(t) = Ax(t)$$

- Invertible

-

$$y(t) = Ax(t) + B$$

- Invertible

-

$$y(n) = x(-n)$$

- Invertible

-

$$y(t) = e^{3x(t)}$$

-

$$y(t) = x^4(t)$$

- Not-invertible

-

$$y(t) = Ax(t)$$

- Invertible

-

$$y(t) = Ax(t) + B$$

- Invertible

-

$$y(n) = x(-n)$$

- Invertible

-

$$y(t) = e^{3x(t)}$$

- Invertible

A continuous-time system is described by the following input-output relationship:

$$y(t) = [\sin 6t]x(t)$$

Determine whether this system is Memoryless, Time invariant, Linear, Causal and Stable?

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Ripal Patel

System
Classification
and properties

System Classification

Memory vs.
MemorylessCausal vs.
Non-causalLinear vs.
NonlinearTime-invariant
vs.
Time-variantStable Vs
Non-stable

Invertibility

Time domain
representation
of LTI SystemLinear time-invariant
systems (LTI
systems)Impulse Response
Convolution Sum

A Discrete-time system is described by the following input-output relationship:

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Ripal Patel

System
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Convolution Sum

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- Used to predict long-term behavior in a system
- The behavior of an LTI system is completely defined by its impulse response

The discrete version of impulse function is defined by

$$\delta(n) = \begin{cases} 1, & n = 0 \\ 0, & n \neq 0 \end{cases}$$

The continuous time version of impulse function,

$$\delta(t) = \begin{cases} 1, & t = 0 \\ 0, & t \neq 0 \end{cases}$$

- The impulse response” of a system, $h[n]$, is the output that it produces in response to an impulse input.
Definition: if and only if $x[n] = \delta[n]$ then $y[n] = h[n]$

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Definition: if and only if $x[n] = \delta[n]$ then $y[n] = h[n]$
- Given the system equation, the impulse response can be found out just by feeding $x[n] = \delta[n]$ into the system.

- Consider the system

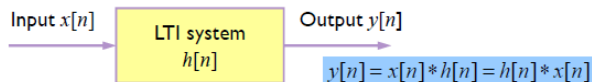
$$y[n] = \frac{1}{2}(x[n] + x[n - 1])$$

- Suppose we insert an impulse:

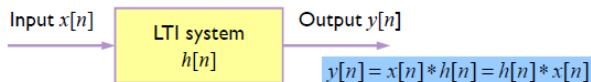
$$x[n] = \delta[n]$$

- Then whatever we get at the output, by Definition, is the impulse response. In this case it is

$$h[n] = \frac{1}{2}(\delta[n] + \delta[n - 1]) = \begin{cases} 0.5, & n = 0, 1 \\ 0, & \textit{otherwise} \end{cases}$$

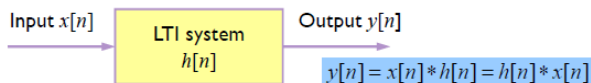


where, $h[n]$ =impulse response of LTI system
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- input-excitation output-response

Find the response $y[n]$ of following LTI system.

$$\mathbf{x(n)=[0,1,2,3,1,0]} \text{ and } \mathbf{h(n)=[0,1,2,2,0]}$$


Ripal Patel

System
Classification
and properties

System Classification

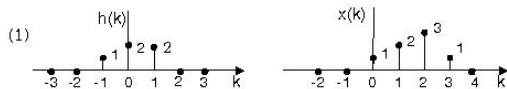
Memory vs.
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Ripal Patel

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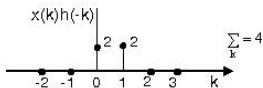
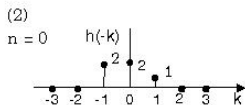
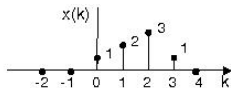
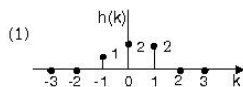
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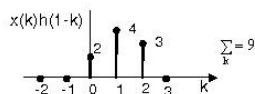
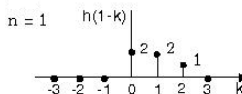
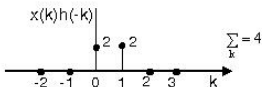
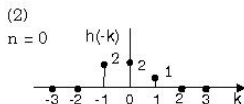
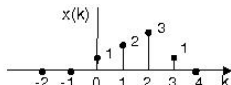
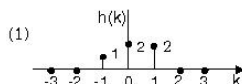
Invertibility

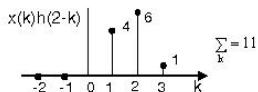
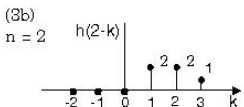
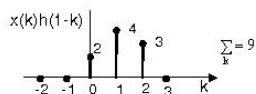
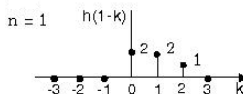
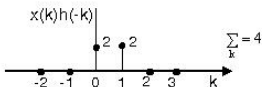
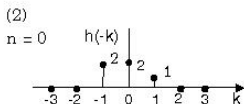
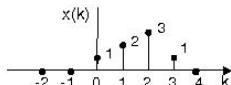
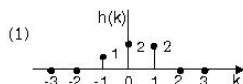
Time domain
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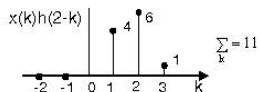
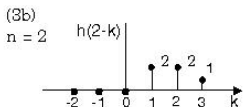
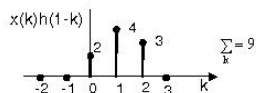
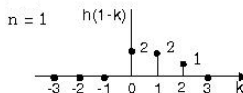
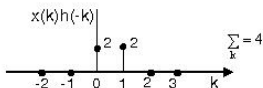
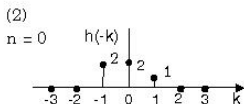
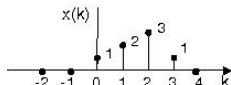
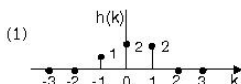
Impulse Response

Convolution Sum









• $y(n) = [\dots, 0, 1, 4, 9, 11, 8, 2, 0, \dots]$

↑

- Size of $x(n)=A=4$, Size of $h(n)=B=3$
Length of $y(n)=A+B-1=4+3-1=6$

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- For $n=-1$

$$y[-1] = \sum_{k=0}^3 x[k]h[-1 - k] =$$

$$x[0]h[-1] + x[1]h[-2] = (1 \times 1) + (2 \times 0) = 1$$

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- Likewise for all the values of n

$$y(n) = [\dots, 0, 1, 4, 9, 11, 8, 2, 0, \dots]$$

↑

$$x_1(n) = [1, 2, 3]$$

$$x_2(n) = [2, 1, 4]$$

$$y[n] = \sum_{k=-\infty}^{\infty} x_1[k]x_2[n-k]$$

- Size of $x_1(n)=A=3$, Size of $x_2(n)=B=3$
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$$y[n] = \sum_{k=-\infty}^{\infty} x_1[k]x_2[n-k]$$

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- For $n=0$

$$\begin{aligned}y[0] &= \sum_{k=0}^2 x_1[k]x_2[-k] = \\&x[0]x_2[0] + x_1[1]x_2[-1] + x_1[2]x_2[-2] \\&= (1 \times 2) + (2 \times 0) + (3 \times 0) = 2\end{aligned}$$

- For $n=0$

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$$= (1 \times 2) + (2 \times 0) + (3 \times 0) = 2$$

- Likewise for all the values of n $y(n) = [2, 5, 12, 11, 12]$

The End