

System Characterization (Chapter 1.6)

This class is focused on analyzing signals and systems and their relationship

We focus on a class of systems called linear systems

- More specifically, we focus on linear time-invariant (LTI) systems
(more on the TI later)

Why?

- Many practical systems (e.g., RLC circuits, op amps, etc.) are linear or approx. linear
- Linear systems have nice properties $\rightarrow$ easy to analyze (compared to non-linear)

Notation: $y(t) = \mathcal{H}\{x(t)\}$

General system model

Systems take in a function as input
and outputs another function
Signals modeled as functions

Suppose $x(t) = a_1 x_1(t) + a_2 x_2(t)$

If and only if a system $\mathcal{H}$ is linear, then

$$y(t) = \mathcal{H}\{a_1 x_1(t) + a_2 x_2(t)\} = a_1 \mathcal{H}\{x_1(t)\} + a_2 \mathcal{H}\{x_2(t)\}$$

Linear system model

Components of Linearity

1. Scaling (or homogeneity)
2. Additivity

Scaling Property: amplitude scaling of input = identical scaling of output

$$\mathcal{H}\{x(t)\} = y(t) \rightarrow \mathcal{H}\{ax(t)\} = ay(t)$$

Ex. $y(t) = \sin(\pi t)x(t)$

Let $x_1(t) = 3x(t) \rightarrow y_1(t) = \sin(\pi t)x_1(t) = 3\sin(\pi t)x(t) = 3y(t)$ ✓

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

Let $x_1(t) = 3x(t) \rightarrow y_1(t) = (x_1(t))^2 = (3x(t))^2 = 9(x(t))^2 \neq 3y(t)$ ✗

Additivity Property: Adding two input signals produces the same output as the sum of the original inputs separately

Let $\mathcal{H}\{x_1(t)\} = y_1(t)$, $\mathcal{H}\{x_2(t)\} = y_2(t) \rightarrow \mathcal{H}\{x_1(t) + x_2(t)\} = y_1(t) + y_2(t)$

Ex. $y(t) = \sin(\pi t)x(t) \rightarrow y_1(t) = \sin(\pi t)x_1(t)$, $y_2(t) = \sin(\pi t)x_2(t)$

Let $x_3(t) = x_1(t) + x_2(t) \rightarrow y_3(t) = \sin(\pi t)(x_1(t) + x_2(t))$
 $= \sin(\pi t)x_1(t) + \sin(\pi t)x_2(t) = y_1(t) + y_2(t)$ ✓

Ex. $y(t) = (x(t))^2 \rightarrow y_1(t) = (x_1(t))^2$, $y_2(t) = (x_2(t))^2$

Let $x_3(t) = x_1(t) + x_2(t) \rightarrow y_3(t) = (x_1(t) + x_2(t))^2 = (x_1(t))^2 + 2x_1(t)x_2(t) + (x_2(t))^2$
 $\neq y_1(t) + y_2(t)$ ✗

• Concepts sometimes referred to as superposition

• Linear systems must satisfy both properties

Proving Linearity:

• Need to show that the above 2 properties hold

• Can do this all at once as well

- Show that if $H\{x_1(t)\} = y_1(t)$, $H\{x_2(t)\} = y_2(t)$,

then
$$H\{a_1 x_1(t) + a_2 x_2(t)\} = a_1 y_1(t) + a_2 y_2(t)$$

- More generally,
$$H\left\{\sum_{k=1}^K a_k x_k(t)\right\} = \sum_{k=1}^K a_k y_k(t)$$

Need to prove this

Can be shown in reverse
also if it is easier to do

Steps:

1. Define $x_1(t)/x_1[n]$ and $x_2(t)/x_2[n]$

$\rightarrow y_1(t) = H\{x_1(t)\}$ and $y_2(t) = H\{x_2(t)\}$

2. Define $x_3(t)/x_3[n]$ as scaled sum

$\rightarrow x_3(t) = a_1 x_1(t) + a_2 x_2(t)$

$\rightarrow y_3(t) = H\{x_3(t)\} \rightarrow$ Scaled & summed at system input

3. Show whether

$y_3(t) = a_1 y_1(t) + a_2 y_2(t) \rightarrow$ Scaled & summed at system output

If equivalent $\rightarrow$ Linear ✓

If not equivalent $\rightarrow$ Not linear ✗

Example: $y[n] = mx[n] + b$

1. Define $x_1[n] \& x_2[n] \rightarrow \underline{y_1[n] = mx_1[n] + b}$
 $\underline{y_2[n] = mx_2[n] + b}$

2. Define $x_3[n] = a_1 x_1[n] + a_2 x_2[n]$
 $\rightarrow \underline{y_3[n] = mx_3[n] + b}$

3. Does $\underline{y_3[n] = a_1 y_1[n] + a_2 y_2[n]}$?

$$a_1 y_1[n] + a_2 y_2[n] = a_1 m x_1[n] + a_1 b + a_2 m x_2[n] + a_2 b$$

$$y_3[n] = mx_3[n] + b = m(a_1 x_1[n] + a_2 x_2[n]) + b$$
$$= a_1 m x_1[n] + a_2 m x_2[n] + b$$

Not equal

$\rightarrow$ Not linear

Example: $V_i(t) = V_o(t) + RC \frac{dv_o(t)}{dt}$

$$V_o(t) = \mathcal{H}(\{v_i(t)\})$$

1. Define $y_1(t) = \mathcal{H}(\{x_1(t)\})$ and $y_2(t) = \mathcal{H}(\{x_2(t)\})$

$$\rightarrow \underline{x_1(t) = y_1(t) + RC \frac{dy_1(t)}{dt}}, \quad \underline{x_2(t) = y_2(t) + RC \frac{dy_2(t)}{dt}}$$

2. Define $x_3(t) = a_1 x_1(t) + a_2 x_2(t)$
 $\rightarrow \underline{x_3(t) = y_3(t) + RC \frac{dy_3(t)}{dt}}$

3. Does $\underline{y_3(t) = a_1 y_1(t) + a_2 y_2(t)}$?

This is a little harder - but we can prove the reverse:
i.e., if $\underline{y_3(t) = a_1 y_1(t) + a_2 y_2(t)}$, does $\underline{x_3(t) = a_1 x_1(t) + a_2 x_2(t)}$?

$$\text{Let } y(t) = a_1 y_1(t) + a_2 y_2(t)$$

$$\text{Then } x(t) = y(t) + RC \frac{dy(t)}{dt} = a_1 y_1(t) + a_2 y_2(t) + \underbrace{RC \frac{d}{dt} (a_1 y_1(t) + a_2 y_2(t))}_{\text{separate}}$$

$$= a_1 y_1(t) + a_2 y_2(t) + a_1 RC \frac{dy_1(t)}{dt} + a_2 RC \frac{dy_2(t)}{dt}$$

$$= a_1 \left(y_1(t) + RC \frac{dy_1(t)}{dt} \right) + a_2 \left(y_2(t) + RC \frac{dy_2(t)}{dt} \right)$$

$$= a_1 x_1(t) + a_2 x_2(t) = \underline{x_3(t)}$$

→ A superposition at the output implies the same superposition at the input

→ Linear

Note: ANY differential equation w/ constant coefficients in a linear form defines a linear system (e.g., any RLC circuit)

$$\text{Example: } \frac{dx(t)}{dt} + 3x(t) = \frac{d^2y(t)}{dt^2} + 4y(t) \rightarrow \text{linear}$$

$$\text{Example: } x(t) \frac{dx(t)}{dt} = \frac{d^2y(t)}{dt^2} \rightarrow \text{Not linear}$$

There is an analogous equation in DT called a difference equation which is also always linear → more on this later in the semester

Time-Invariance

• LTI systems are linear and time-invariant

A system is time-invariant if a delay in the input by t_0 seconds or n_0 samples produces the same output signal, just delayed by t_0 seconds or n_0 samples

- Time shift at input produces identical time shift at output
- I/O characteristics do not change w/ time

Mathematically: $\mathcal{H}\{x(t)\} = y(t) \rightarrow \mathcal{H}\{x(t-t_0)\} = y(t-t_0)$
for all $x(t)$ and t_0

Proving Time-Invariance

1. Define $x_1(t)$ w/ $y_1(t) = \mathcal{H}\{x_1(t)\}$
2. Define $x_2(t) = x_1(t-t_0) \rightarrow$ $y_2(t) = \mathcal{H}\{x_2(t)\}$
3. Show whether $y_1(t-t_0) = y_2(t)$

If equivalent $\rightarrow$ Time-invariant $\checkmark$
If not $\rightarrow$ Time-varying $\times$

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

1. Define $x_1(t) \rightarrow$ $y_1(t) = x_1(t^2)$
2. Define $x_2(t) = x_1(t-t_0) \rightarrow$ $y_2(t) = \mathcal{H}\{x_2(t)\}$
3. Does $y_1(t-t_0) = y_2(t)$?

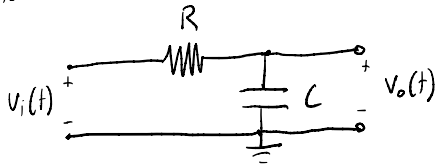
$$y_2(t) = x_2(t^2) = x_1(t^2 - t_0)$$

$$y_1(t-t_0) = x_1((t-t_0)^2) = x_1(t^2 - 2t t_0 + t_0^2)$$

Not equal $\rightarrow$ Time-varying

Why LTI?

Consider the basic RC circuit:



$$\rightarrow v_o(t) = \frac{\frac{1}{j\omega C}}{R + \frac{1}{j\omega C}} v_i(t) = \underbrace{\frac{1}{1 + j\omega RC}}_{\text{Defined for specific values of } \omega} v_i(t) = H\{v_i(t)\}$$

Suppose $R = 500 \Omega$ and $C = 5 \text{ nF}$; $v_i(t) = e^{j\omega t}$ Only one ω value

• At $\omega = 0 \text{ rad/s} \rightarrow v_i(t) = 1 \text{ (DC)} \rightarrow v_o(t) = \frac{1}{1 + j(0)RC} (1) = 1 = v_i(t) \rightarrow \text{Unchanged}$

• At $\omega = 400,000 \text{ rad/s} \rightarrow v_i(t) = e^{j400,000t} \rightarrow v_o(t) = \frac{1}{1 + j400,000RC} e^{j400,000t}$
 $= 0.707 e^{j400,000t} e^{-j\pi/4} = 0.707 e^{-j\pi/4} v_i(t)$
Original input scaled by half power & 45° phase shift

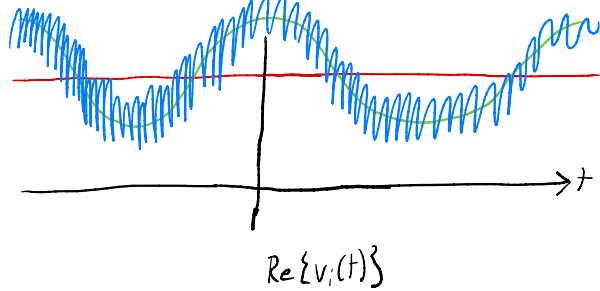
• At $\omega = 40 \text{ Mrad/s} \rightarrow v_i(t) = e^{j(40 \cdot 10^6)t} \rightarrow v_o(t) = \frac{1}{1 + j(40 \cdot 10^6)RC} e^{j(40 \cdot 10^6)t}$
 $\approx 0.01 e^{j(40 \cdot 10^6)t} e^{-j\frac{\pi}{2}} = 0.01 e^{-j\frac{\pi}{2}} v_i(t)$
Original input scaled by 10^{-4} in power & 90° phase shift

Notice:

- Each input frequency produces an output sinusoid at the same frequency
- Each frequency is scaled with a different amplitude and phase shift (a complex scale factor) defined as a function of $\omega \rightarrow$ call this $H(\omega)$

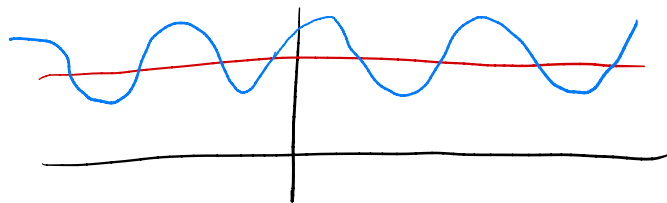
What if input is multiple frequencies?

e.g., $v_i(t) = 1 + e^{j400,000t} + e^{j(40 \cdot 10^4)t}$



Remember linearity! Scale & phase shift each sinusoid separately, then add back together:

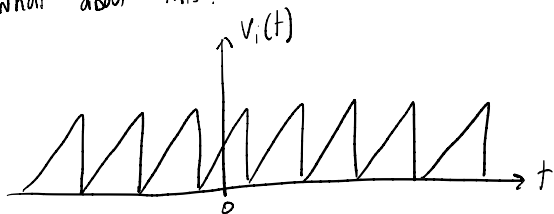
$$v_o(t) = 1 + 0.707e^{-j\pi/4}e^{j400,000t} + 0.01e^{-j\pi/2}e^{j(40 \cdot 10^4)t}$$



So, $v_o(t) = \frac{1}{1 + j\omega RL} v_i(t)$

Frequency-dependent
scale factor $H(\omega)$

What about this?



???

• Complicated signal, no obvious discrete sinusoids...

• It turns out we can apply a transform to $v_i(t)$ to get

$$v_i(t) = \underline{a_1 e^{j\omega_1 t}} + \underline{a_2 e^{j\omega_2 t}} + \underline{a_3 e^{j\omega_3 t}} + \dots$$

So at the output we can compute:

$$v_o(t) = \underline{a_1 H(\omega_1) e^{j\omega_1 t}} + \underline{a_2 H(\omega_2) e^{j\omega_2 t}} + \underline{a_3 H(\omega_3) e^{j\omega_3 t}} + \dots$$

Each frequency comes out of the LTI system as the same frequency, just scaled and phase shifted by $H(\omega)$ independently of all other frequencies

Like eigenvectors in matrix algebra $Ax = \lambda x$, but now $H\{e^{j\omega_1 t}\} = H(\omega_1) e^{j\omega_1 t}$

This fundamental relationship between LTI systems and sinusoids makes LTI systems easier to work with and sinusoids convenient for constructing signals.

The transform we applied above is called the continuous-time Fourier Series.

In this class we will learn the following transforms:

- CT and DT periodic signals: Fourier Series
- CT aperiodic signals: Fourier (Ω or f) and Laplace (s) Transforms
- DT aperiodic signals: Discrete-Time Fourier (ω or f) and z - (z) Transforms

Causality

A system is causal if output at $t=t_0$ only depends on input for $t \leq t_0$

- System output doesn't require prediction of future values
- A requirement for real-time systems

Examples

$$y(t) = -x(t+2) + x(t+3) \rightarrow \text{Try } t=0 \rightarrow \text{requires } x(3) \rightarrow \underline{\text{Non-causal}}$$

$$y[n] = \sum_{k=0}^M a_k x[n-k] \rightarrow \text{Try } n=0 \rightarrow \text{requires } x[0], \dots, x[M] \rightarrow \underline{\text{Causal}}$$

$$y[n] = \sum_{k=-1}^M a_k x[n-k] \rightarrow \text{Try } n=0 \rightarrow \text{requires } x[1] \rightarrow \underline{\text{Non-causal}}$$

Memoryless vs. With Memory

- Memoryless systems have outputs at each value of t or n that only depend on values of the input at that time
 - Resistor: $v(t) = R i(t) \rightarrow$ Memoryless
 - Capacitor: $v(t) = \frac{1}{C} \int_{-\infty}^t i(\tau) d\tau \rightarrow$ has (infinite) memory

Invertible Systems

- In invertible systems, each unique input produces a unique output
- If a system is invertible, there exists an "inverse" system that produces the original input signal from the output signal
 - i.e., if $\mathcal{H}\{x(t)\} = y(t)$, then $\mathcal{H}^{-1}\{y(t)\} = x(t)$
- Example: $y(t) = (x(t))^2$ is not invertible

Bounded-Input, Bounded-Output (BIBO) Stable

- If the input to a BIBO stable system is bounded/finite, the output will also be bounded/finite (no divergence)
- Unstable example: $y[n] = \sum_{k=0}^{\infty} x[n-k]$ (Try $x[n] = u[n]$)

LTI Example:

Suppose that for some system H , when the input $x(t) = \delta(t)$, the output $y(t) = e^{-t}u(t)$.

Now, when the input is $x_1(t) = 3\delta(t) - 2\delta(t-2)$, find the output $y_1(t)$.

Linearity: $Ax(t) \xrightarrow{H} Ay(t)$ and $x_1(t) + x_2(t) \xrightarrow{H} y_1(t) + y_2(t)$

We can find the response to each $\delta(t)$ and add together results!

Also, the scaling factors follow through an LTI system.

$$\rightarrow 3\delta(t) \xrightarrow{H} 3e^{-t}u(t)$$

What about the time-shift in $-2\delta(t-2)$?

Time-Invariance: $x(t-t_0) \xrightarrow{H} y(t-t_0)$

We can use the same $y(t)$, just shifted!

$$\rightarrow -2\delta(t-2) \xrightarrow{H} -2e^{-(t-2)}u(t-2)$$

$$\rightarrow y_1(t) = 3e^{-t}u(t) - 2e^{-(t-2)}u(t-2)$$

Another way: $x(t) = \delta(t) \xrightarrow{H} y(t) = e^{-t}u(t) \rightarrow x_1(t) = 3\delta(t) - 2\delta(t-2) = 3x(t) - 2x(t-2)$

$$\xrightarrow{H} y_1(t) = 3y(t) - 2y(t-2) = 3e^{-t}u(t) - 2e^{-(t-2)}u(t-2)$$

Another Example: $x(t) = 4\delta(t-1) \rightarrow \boxed{H} \rightarrow y(t) = 8\cos\left(\frac{\pi}{2}(t-2)\right)u(t-2)$

Find $y_1(t)$ when $x_1(t) = 3\delta(t+1)$.

First: work backward to find $y_2(t)$ when $x_2(t) = \delta(t)$

$$x(t) = 4\delta(t-1) \xrightarrow{H} y(t) = 8\cos\left(\frac{\pi}{2}(t-2)\right)u(t-2) \rightarrow x_2(t) = \delta(t) \xrightarrow{H} y_2(t) = \frac{1}{4}y(t+1) \\ = \underline{2\cos\left(\frac{\pi}{2}(t+1)\right)u(t+1) = y_2(t)}$$

$$\rightarrow \text{So: } x_1(t) = 3\delta(t+1) \xrightarrow{H} y_1(t) = 3y_2(t+1) = 6\cos\left(\frac{\pi}{2}t\right)u(t)$$