

EE 123 Discussion Section 1

https://xkcd.com/26/

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Based on notes by Suma Anand, Josh Sanz, Li-Hao Yeh, Jon Tamir, Giulia Fanti and Frank Ong

Announcements

- HW 1 due Friday, Jan 27
- Check website for OH calendar
- Lab 0 due next week, Lab 1 in 3 weeks
- Discussion guidelines:
 - Respect fellow students and staff
 - We are here to learn
 - Wrong answers ok
 - Questions encouraged!

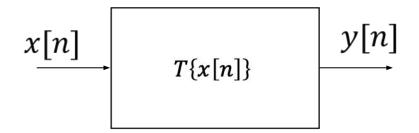
Today

- Properties of systems
- Review of linear regression (least-squares)

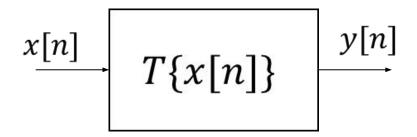
Concept Review

- Causal
 - $y[n_0] = f(x[n], x[n-1], ...), -\infty < n \le n_0$
- Memoryless
 - $y[n] = f(x[n]) \ \forall n$
- Linear
 - $T\{x_i[n]\}=y_i[n]$
 - $T\{\alpha x_1[n] + \beta x_2[n]\} = \alpha y_1[n] + \beta y_2[n]$
- Time Invariant
 - $T\{x[n-n_0]\} = y[n-n_0]$
- BIBO Stable
 - $|x[n]| \le B_x < \infty \Rightarrow |y[n]| \le B_y < \infty \ \forall \ n$
 - LTI: stable IFF $\sum |h[k]| < \infty$

Concept Review



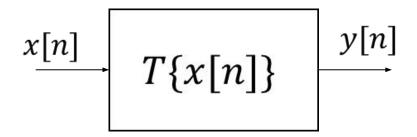
- Impulse response
 - $h[n] = T\{\delta[n]\}$
- Convolution
 - $y[n] = h[n] * x[n] = \sum_{k=-\infty}^{\infty} h[k]x[n-k]$



Consider a system below:

$$y[n] = \alpha x[n]$$

Is this system Linear/Time-invariant/Causal/BIBO stable?

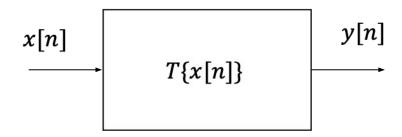


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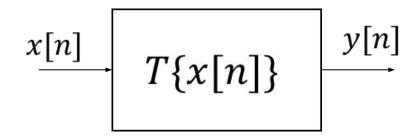
YES to all



What about this modified system:

$$y[n] = \begin{cases} \alpha x[n], & x[n] \le 1 \\ \alpha, & x[n] > 1 \end{cases}$$

Is this system Linear/Time-invariant/Causal/BIBO stable?



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$$y[n] = \begin{cases} \alpha x[n], & x[n] \le 1 \\ \alpha, & x[n] > 1 \end{cases}$$

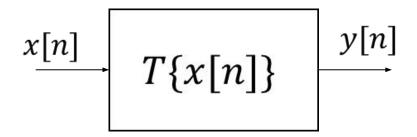
Is this system Linear/Time-invariant/Causal/BIBO stable?

Not linear \rightarrow whenever x[n] > 1, $y[n] = \alpha$, this is not homogeneous.

Time invariant \rightarrow Plug in x[n-1], the output is y[n-1]

Causal \rightarrow It only depends on current x[n], which is memoryless/causal

Stable \rightarrow both α and x[n] are bounded then output y[n] is bounded



What about this modified system:

$$y[n] = \begin{cases} \alpha x[n], & x[n] \le 1 \\ \alpha, & x[n] > 1 \end{cases}$$

Do real systems act this way?

- 1. MOS/BJT amplifier input/output
- 2. Hooke's law (elastic materials)
- 3. Hysteresis

They all have a limited linear region

Another system (from old exam)

A discrete-time system H produces an output signal y that is the symmetric part of the input:

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

Which of the following are true?

- The system must be LTI
- The system cannot be LTI

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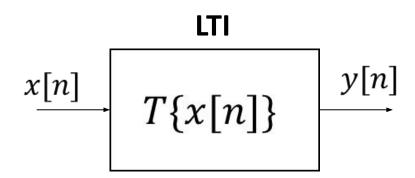
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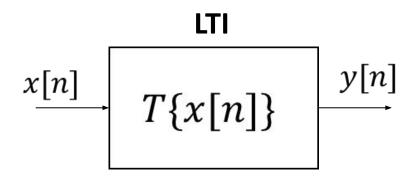
Not time invariant:

- For $x_1[n] = \delta[n]$, then $y_1[n] = \delta[n]$
- For $x_2[n] = \delta[n-1]$, then $y_2[n] = \frac{\delta[n-1] + \delta[n+1]}{2}$
- $y_1[0] = 1$ but $y_2[1] = \frac{1}{2}$
- → Not time invariant

(however, the system is linear)



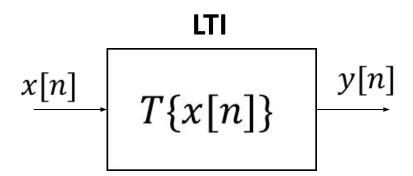
Consider an LTI system with input x[n] and output y[n]. When we input a signal $\left(\frac{1}{3}\right)^n u[n]$, where u[n] is unit step function, we observe an output g[n]. Can we express y[n] in terms of x[n] and g[n]?



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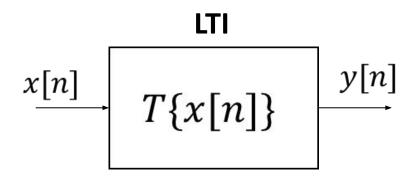
The key is to massage the input into $oldsymbol{\delta}[oldsymbol{n}]$

We know
$$u[n] - u[n-1] = \delta[n]$$



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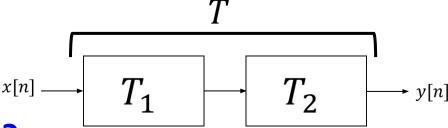
By time-invariance: input $\left(\frac{1}{3}\right)^{n-1}u[n-1]$, output g[n-1] By linearity: input $\left(\frac{1}{3}\right)^n(u[n]-u[n-1])=\delta[n]$, output $g[n]-\frac{1}{3}g[n-1]$ (impulse response) $y[n]=x[n]*\left(g[n]-\frac{1}{3}g[n-1]\right)$



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Cascaded system problem

Let T1 and T2 be two separate systems and T be the cascaded system:



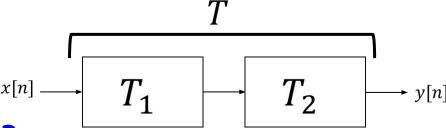
True or False?

If T1 is LTI and T2 is not LTI, then T cannot be LTI

If T1 is not LTI and T2 is not LTI, then T cannot be LTI

Cascaded system problem

Let T1 and T2 be two separate systems and T be the cascaded system:



True or False?

If T1 is LTI and T2 is not LTI, then T cannot be LTI

False

Consider the system T1=0. Then T=0

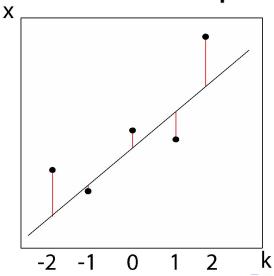
If T1 is not LTI and T2 is not LTI, then T cannot be LTI

False

Consider the system
$$T_1\{x\} = x^3$$
 and $T_2\{x\} = x^{\frac{1}{3}}$. Then $T\{x\} = x$

Many signal processing problems can be formulated as a **least squares** problem, where we try to find model parameters that best fit the observed data. We will see this many, many times

Example: Linear regression. Suppose we observe five data points x[k], where $k = \{-2, -1, 0, 1, 2\}$. We want to fit a line x = mk + b by minimizing the squared distance between the line and the data points:



- 1. We want to write the squared distance in the form $\frac{1}{2} ||\mathbf{x} \mathbf{K}\boldsymbol{\beta}||_2^2$, where $\mathbf{x} = [x[-2], x[-1], ..., x[2]]^T$ and $\boldsymbol{\beta} = [m\ b]^T$. What is K?
- Solve for m and b in terms of K and x.

For each value of k, we have a linear equation for our model:

Example,
$$k = 2$$
: $x[2] = 2m + b$

And we have a squared error with our data:

Example,
$$k = 2: (x[2] - (b + 2m))^2$$

Sum of squared errors: $\sum_{k} (x[k] - (mk + b))^{2}$

⇒ In matrix form, Error =
$$\frac{1}{2} \|\mathbf{x} - \mathbf{K}\boldsymbol{\beta}\|_2^2$$

Error =
$$\frac{1}{2} \left\| \begin{pmatrix} x_{-2} \\ x_{-1} \\ x_0 \\ x_1 \\ x_2 \end{pmatrix} - \begin{bmatrix} -2 & 1 \\ -1 & 1 \\ 0 & 1 \\ 1 & 1 \\ 2 & 1 \end{bmatrix} \begin{pmatrix} m \\ b \end{pmatrix} \right\|_{2}^{2}$$

To find the best fit from a least squares sense, minimize the sum of squared errors:

$$\underset{m,b}{\text{minimize}} \frac{1}{2} \left\| \begin{pmatrix} x_{-2} \\ x_{-1} \\ x_0 \\ x_1 \\ x_2 \end{pmatrix} - \begin{bmatrix} -2 & 1 \\ -1 & 1 \\ 0 & 1 \\ 1 & 1 \\ 2 & 1 \end{bmatrix} {m \choose b} \right\|_{2}^{2} = \underset{\beta}{\text{minimize}} \frac{1}{2} \|\mathbf{x} - \mathbf{K}\boldsymbol{\beta}\|_{2}^{2}$$

To solve for b and m, take the derivative (gradient) with respect to b and to m, and set to zero:

$$\underset{\beta}{\text{minimize}} \frac{1}{2} \|\mathbf{x} - \mathbf{K}\boldsymbol{\beta}\|_{2}^{2}$$

$$\mathbf{K}^{\mathsf{T}}\mathbf{K}\boldsymbol{\beta} - \mathbf{K}^{\mathsf{T}}\mathbf{x} = 0 \implies \boldsymbol{\beta} = (\mathbf{K}^{\mathsf{T}}\mathbf{K})^{-1}\mathbf{K}^{\mathsf{T}}\mathbf{x}$$

In Python,

```
K = np.array( [...] )
x = np.array( [...])
beta = np.linalg.solve(K, x)
```