

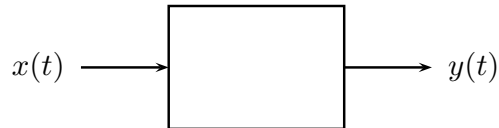
EE120 - Fall'15 - Lecture 1 Notes¹

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Linear Time-Invariant (LTI) Systems



Linearity: Two conditions must be satisfied:

1. Scaling:

$$ax(t) \rightarrow ay(t) \text{ for any number } a; \quad (1)$$

2. Superposition:

$$x_1(t) + x_2(t) \rightarrow y_1(t) + y_2(t). \quad (2)$$

Corollary: If the input to a linear system is 0, the output must be 0.

Proof. Choose $a = 0$ in the scaling property.

Time-Invariance: A time shift in the input results in an identical time shift in the output:

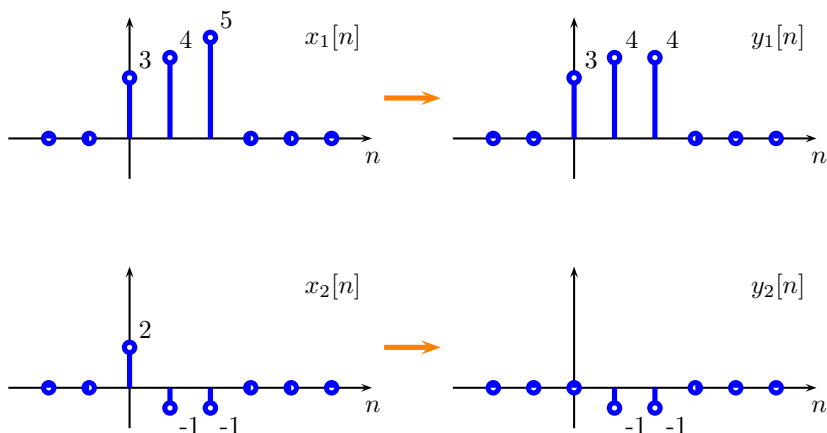
$$x(t - T) \rightarrow y(t - T). \quad (3)$$

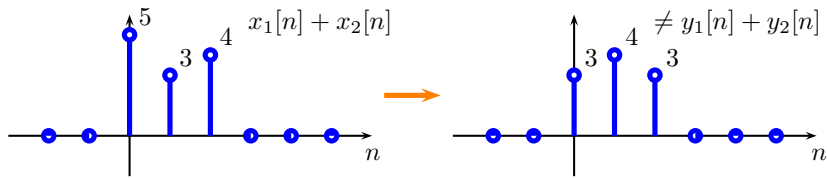
Example: Moving average filter:

$$y[n] = \frac{1}{3} (x[n - 1] + x[n] + x[n + 1]) \rightarrow \text{LTI}. \quad (4)$$

Example: Median Filter:

$$y[n] = \text{med}\{x[n - 1], x[n], x[n + 1]\} \rightarrow \text{TI, but nonlinear}. \quad (5)$$

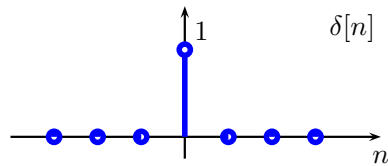




Discrete-Time (DT) LTI Systems: Convolution Sum

Section 2.1 in Oppenheim & Willsky

Let $h[n]$ denote the response of an LTI system to the unit impulse:



Then, for any input $x[n]$, the output is:

$$y[n] = \sum_{k=-\infty}^{\infty} x[k]h[n-k] \quad \text{"convolution sum"} \quad (6)$$

Proof. Rewrite $x[n]$ as

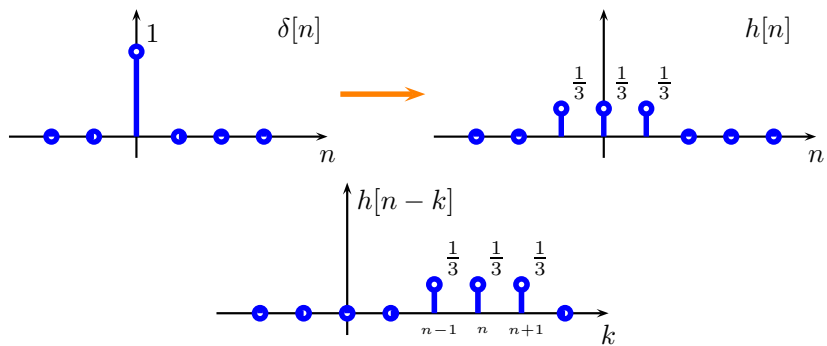
$$x[n] = \dots + x[-1]\delta[n+1] + x[0]\delta[n] + x[1]\delta[n-1] + \dots \quad (7)$$

$$= \sum_{k=-\infty}^{\infty} x[k]\delta[n-k] \quad (8)$$

Since $\delta[n] \rightarrow h[n]$, by time-invariance: $\delta[n-k] \rightarrow h[n-k]$.

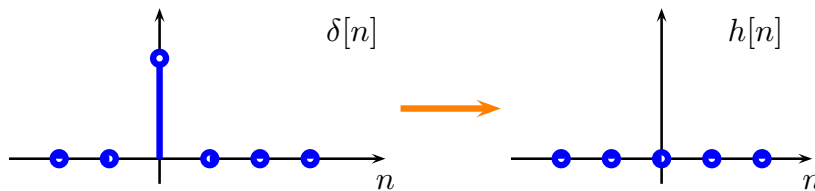
Then, by linearity: $\sum_k x[k]\delta[n-k] \rightarrow \sum_k x[k]h[n-k]$.

Example: For the moving average system above,



$$\begin{aligned}
 y[n] &= \sum_{k=-\infty}^{\infty} x[k]h[n-k] \\
 &= \sum_{k=n-1}^{n+1} \frac{1}{3}x[k] = \frac{1}{3}(x[n-1] + x[n] + x[n+1]) \quad (9)
 \end{aligned}$$

Example: For the median filter:



Since the system is nonlinear, we can't use convolution to predict the output.

Continuous-Time (CT) LTI Systems: Convolution Integral

Unit impulse:

$$\delta(t) \triangleq \lim_{\Delta \rightarrow 0} \delta_{\Delta}(t) \quad (10)$$

where $\delta_{\Delta}(t)$ is as in Figure 1.

Let $h(t)$ denote the response of a LTI system to $\delta(t)$.

Then, for any input $x(t)$, the output is :

$$y(t) = \int_{-\infty}^{\infty} x(\tau)h(t-\tau)d\tau \quad \text{"convolution integral"} \quad (11)$$

Proof. First, note that the staircase approximation in Figure 2 recovers $x(t)$ as $\Delta \rightarrow 0$:

$$x(t) = \lim_{\Delta \rightarrow 0} \sum_{k=-\infty}^{\infty} x(k\Delta)\Delta\delta_{\Delta}(t-k\Delta). \quad (12)$$

Next, let $h_{\Delta}(t)$ denote the response of the system to $\delta_{\Delta}(t)$ and note from the LTI property that the response to each term in the sum above is $x(k\Delta)\Delta h_{\Delta}(t-k\Delta)$. Thus, the response to $x(t)$ is

$$y(t) = \lim_{\Delta \rightarrow 0} \sum_{k=-\infty}^{\infty} x(k\Delta)h_{\Delta}(t-k\Delta)\Delta = \int_{-\infty}^{\infty} x(\tau)h(t-\tau)d\tau. \quad (13)$$

Section 2.2 in Oppenheim & Willsky

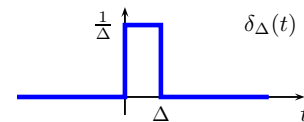


Figure 1: $\delta_{\Delta}(t)$

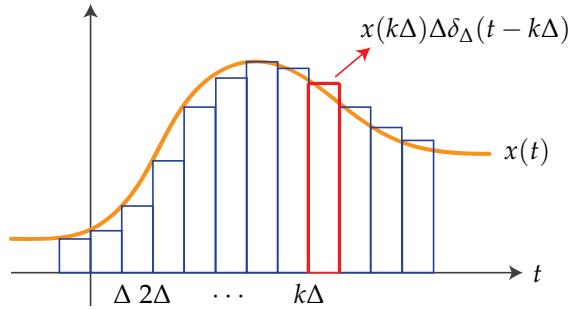


Figure 2: Staircase approximation of $x(t)$.

Properties of LTI Systems

Section 2.3 in Oppenheim & Willsky

We will denote the convolution operation by " $*$ ".

1. Commutative Property:

$$x[n] * h[n] = h[n] * x[n] \tag{14}$$

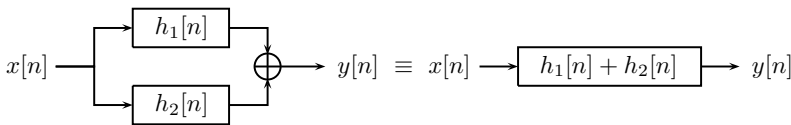
Proof.

$$\sum_{k=-\infty}^{\infty} x[k]h[n-k] = \sum_{r=-\infty}^{\infty} x[n-r]h[r], \tag{15}$$

with the change of variables $(n - k) \triangleq r$.

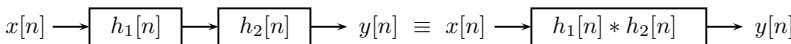
2. Distributive Property:

$$x[n] * (h_1[n] + h_2[n]) = x[n] * h_1[n] + x[n] * h_2[n] \tag{16}$$



3. Associative Property:

$$x[n] * (h_1[n] * h_2[n]) = (x[n] * h_1[n]) * h_2[n] \tag{17}$$



Combine this with the commutative property



Properties 1,2,3 above also hold for CT systems.

Determining Causality from the Impulse Response

For a DT LTI system, causality means:

$$h[n] = 0, \quad \forall n < 0. \quad (18)$$

For a CT LTI system, causality means:

$$h(t) = 0, \quad \forall t < 0. \quad (19)$$

Proof. Since $y[n] = \sum_{k=-\infty}^{\infty} h[k]x[n-k]$, if $h[k] \neq 0$ for some $k < 0$, then $y[n]$ depends on $x[n-k]$, where $n-k > n$.

Example: Moving average system above: $h[-1] \neq 0 \rightarrow$ noncausal.

Determining Stability from the Impulse Response

Stability criterion for a DT LTI system:

$$\sum_{k=-\infty}^{\infty} |h[k]| < \infty. \quad (20)$$

Stability criterion for a CT LTI system:

$$\int_{-\infty}^{\infty} |h(\tau)| d\tau < \infty. \quad (21)$$

Proof.

Sufficiency: Suppose $\sum_{k=-\infty}^{\infty} |h[k]| < \infty$ and show that bounded inputs give bounded outputs:

$$|x[n]| \leq B \text{ for all } n, \text{ for some } B > 0.$$

$$|y[n]| = \left| \sum_k x[n-k]h[k] \right| \leq \sum_k |x[n-k]| \cdot |h[k]| \leq B \sum_k |h[k]| < \infty.$$

Necessity: To prove "stable $\Rightarrow \sum_k |h[k]| < \infty$ " prove the contrapositive:

$$\text{"} \sum_k |h[k]| = \infty \Rightarrow \text{unstable.} \text{"} \quad (22)$$

Let $x[n] = \text{sgn}\{h[-n]\}$. Then, since $y[n] = \sum_{k=-\infty}^{\infty} h[k]x[n-k]$:

$$y[0] = \sum_{k=-\infty}^{\infty} h[k]x[-k] = \sum_k h[k] \text{sign}\{h[k]\} = \sum_k |h[k]| = \infty. \quad (23)$$

Examples:

1. Moving average system above:

$$\sum_k |h[k]| = \frac{1}{3} + \frac{1}{3} + \frac{1}{3} = 1 \rightarrow \text{stable.} \quad (24)$$

2. Integrator: $y(t) = \int_{-\infty}^t x(\tau) d\tau$. $h(t)$ is the unit step (see Figure 3),

and

$$\int_{-\infty}^{\infty} |h(\tau)| d\tau = \infty. \quad (25)$$

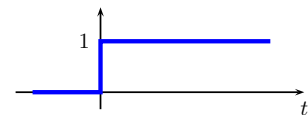


Figure 3: UnitStep