

Signals and Systems

Linear Time-Invariant (LTI) Systems

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Discrete-Time LTI Systems

Representing Signals in Terms of Impulses

- Sifting property

$$x[n] = \sum_{k=-\infty}^{\infty} x[k] \delta[n-k]$$

= ...

$$+ x[-2] \delta[n+2]$$

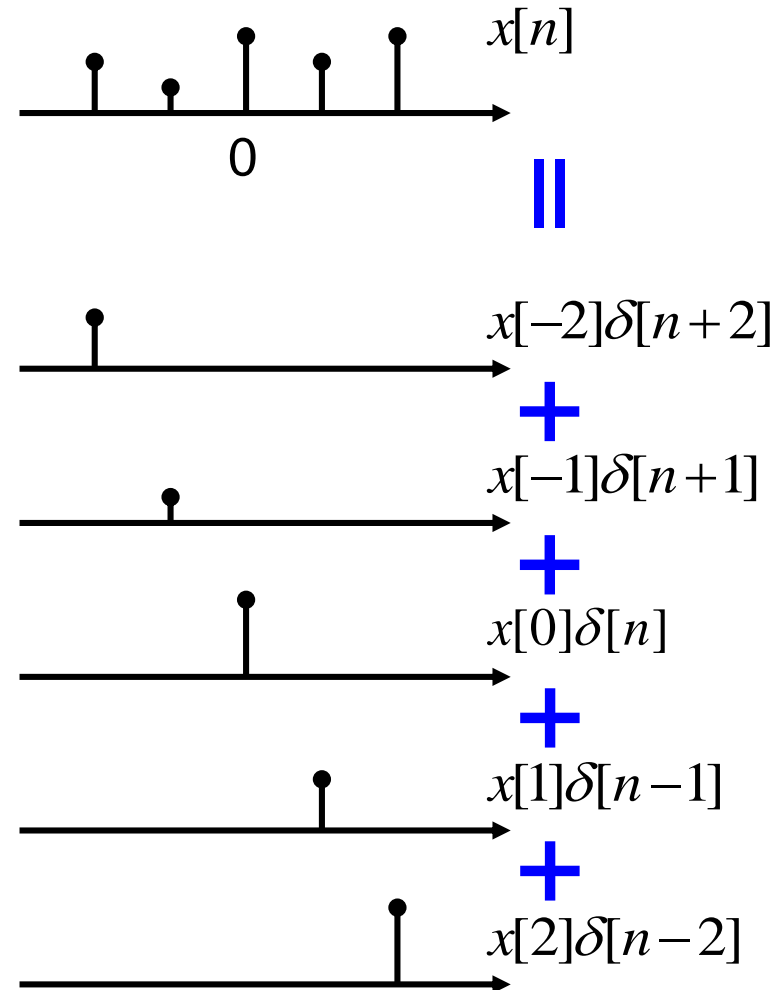
$$+ x[-1] \delta[n+1]$$

$$+ x[0] \delta[n]$$

$$+ x[1] \delta[n-1]$$

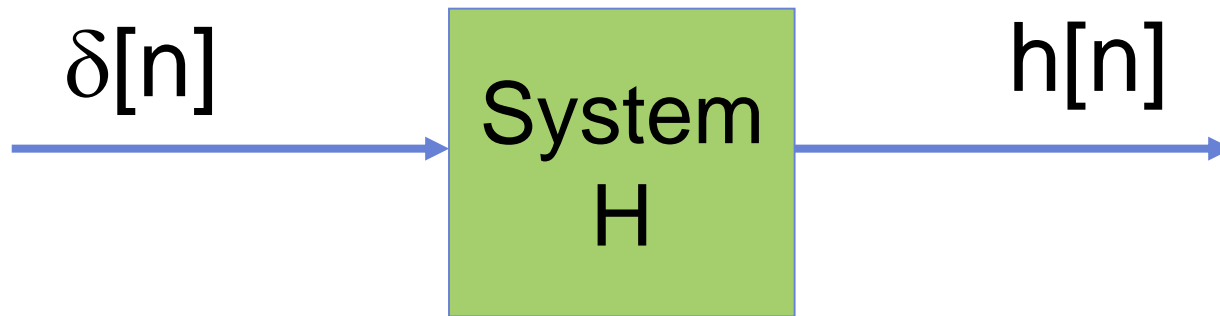
$$+ x[2] \delta[n-2]$$

+ ...



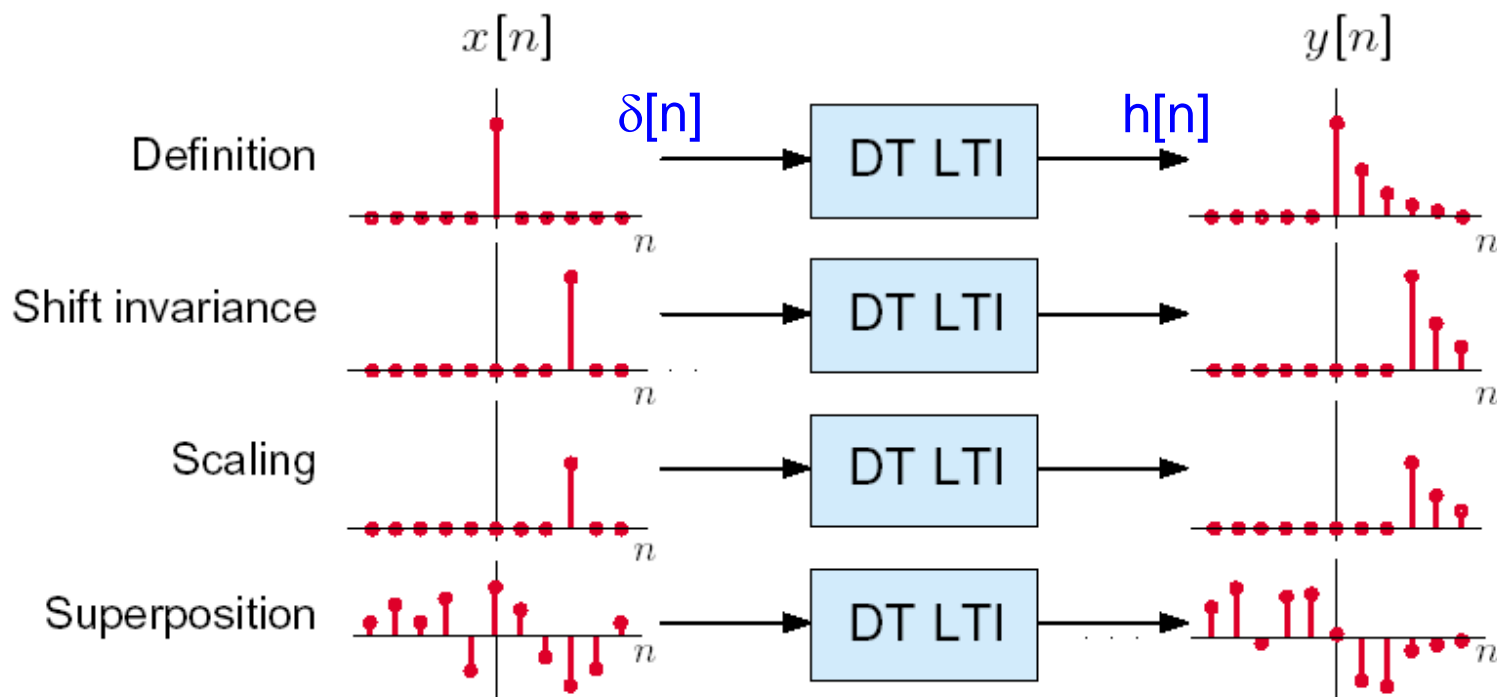
Impulse Response

- The response of a system H to the unit impulse $\delta[n]$ is called the impulse response, which is denoted by $h[n]$
 - ▶ $h[n] = H[\delta[n]]$



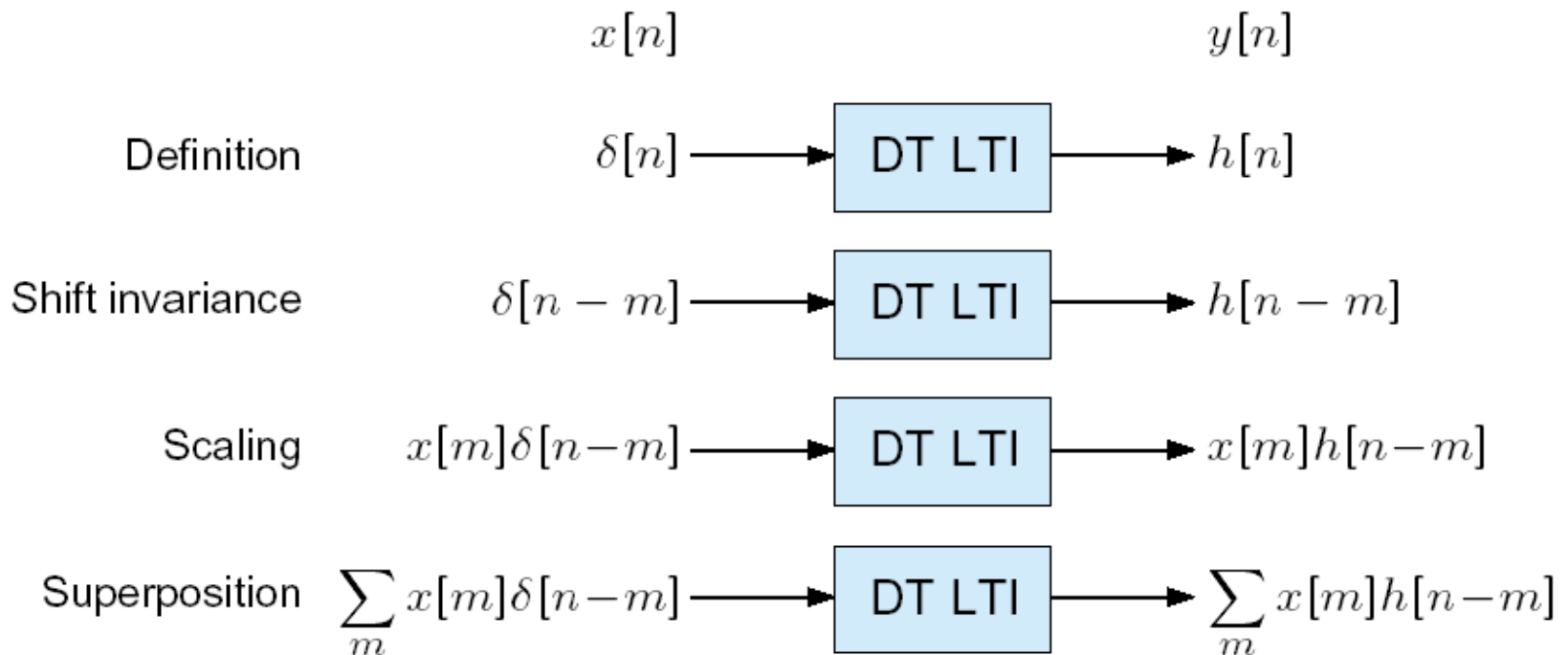
Convolution Sum

- Let $h[n]$ be the impulse response of an LTI system.
- Given $h[n]$, we can compute the response $y[n]$ of the system to any input signal $x[n]$.



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- Given $h[n]$, we can compute the response $y[n]$ of the system to any input signal $x[n]$.

$$x[n] = \sum_{k=-\infty}^{\infty} x[k] \delta[n-k]$$

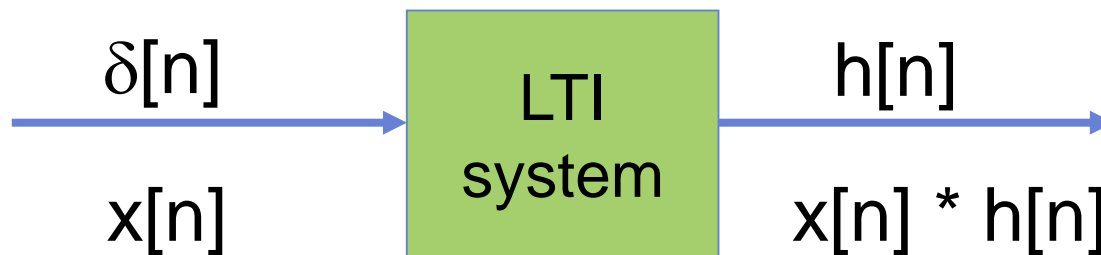
$$\begin{aligned} y[n] &= H[x[n]] \\ &= H \left[\sum_{k=-\infty}^{\infty} x[k] \delta[n-k] \right] \\ &= \sum_{k=-\infty}^{\infty} H [x[k] \delta[n-k]] \\ &= \sum_{k=-\infty}^{\infty} x[k] H [\delta[n-k]] \\ &= \sum_{k=-\infty}^{\infty} x[k] h[n-k] \end{aligned}$$

Convolution Sum

- Notation for convolution sum

$$y[n] = x[n] * h[n] = \sum_{k=-\infty}^{\infty} x[k]h[n-k]$$

- The characteristic of an LTI system is completely determined by its impulse response.



Convolution Sum

- To compute the convolution sum

$$y[n] = x[n] * h[n] = \sum_{k=-\infty}^{\infty} x[k]h[n-k]$$

Step 1 Plot x and h vs k since the convolution sum is on k .

Step 2 Flip $h[k]$ around the vertical axis to obtain $h[-k]$.

Step 3 Shift $h[-k]$ by n to obtain $h[n-k]$.

Step 4 Multiply to obtain $x[k]h[n-k]$.

Step 5 Sum on k to compute $\sum x[k]h[n-k]$.

Step 6 Change n and repeat **Steps 3-6**.

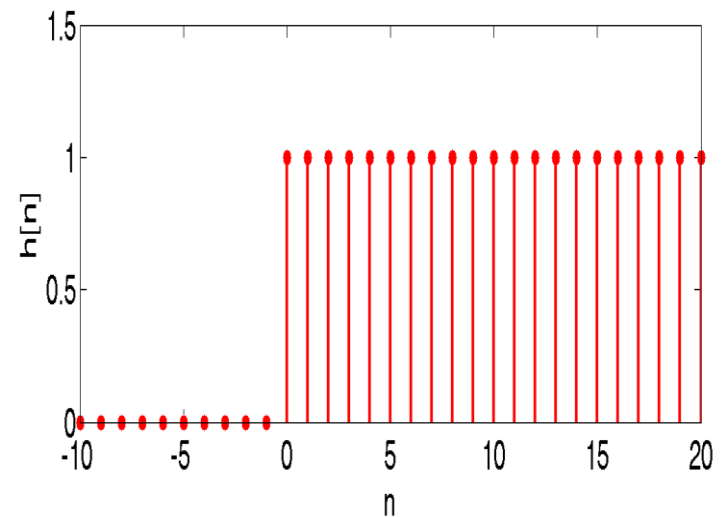
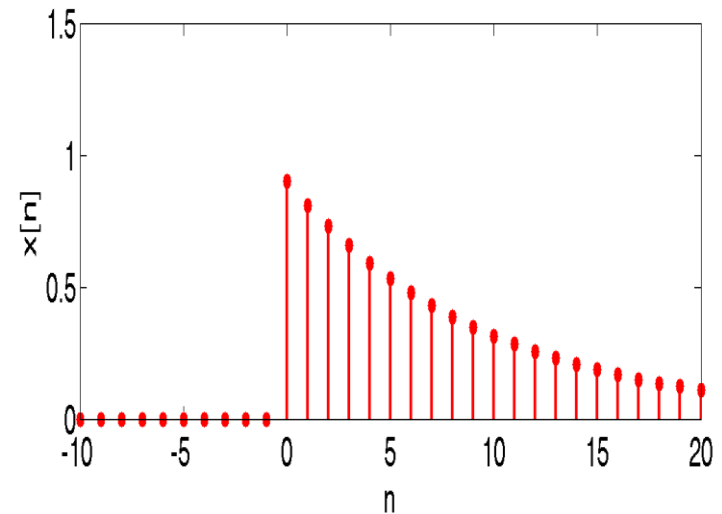
Example

- Consider an LTI system that has an impulse response $h[n] = u[n]$
- What is the response when an input signal is given by
$$x[n] = \alpha^n u[n]$$
where $0 < \alpha < 1$?

- For $n \geq 0$,
$$y[n] = \sum_{k=0}^n \alpha^k$$
$$= \frac{1 - \alpha^{n+1}}{1 - \alpha}$$

- Therefore,

$$y[n] = \left(\frac{1 - \alpha^{n+1}}{1 - \alpha} \right) u[n]$$



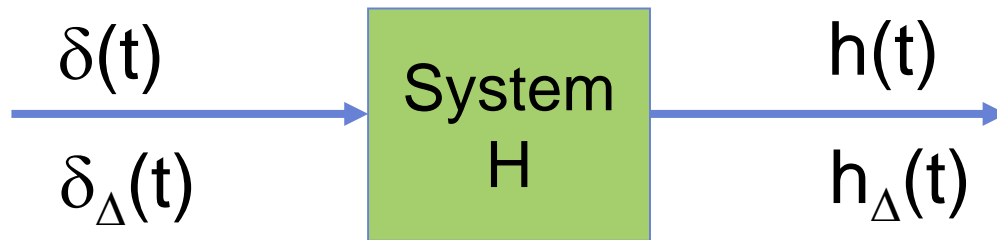
Convolution Sum

Demonstration

Continuous-Time LTI Systems

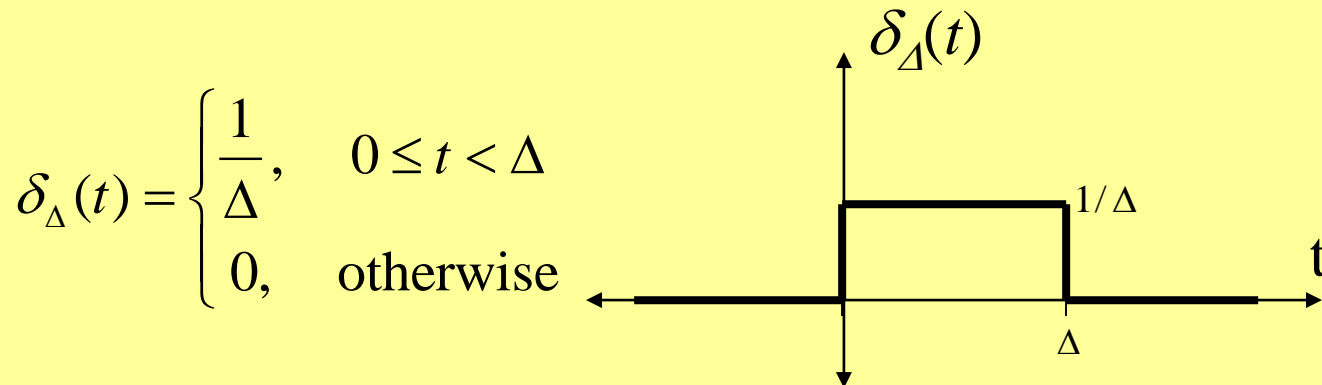
Impulse Response

- The response of a system H to the unit impulse $\delta(t)$ is called the impulse response, which is denoted by $h(t)$
 - ▶ $h(t) = H(\delta(t))$



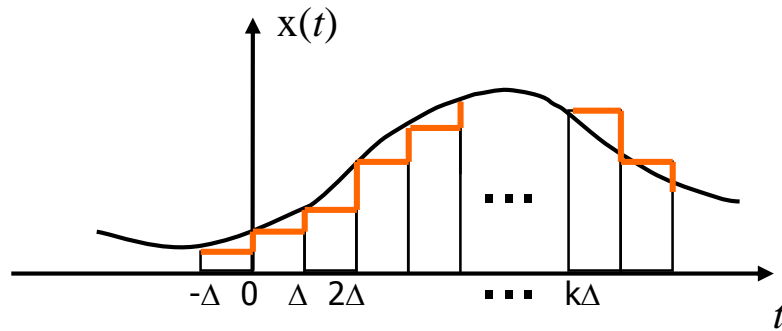
- ▶ As $\Delta \rightarrow 0$, $\delta_{\Delta}(t) \rightarrow \delta(t)$ and $h_{\Delta}(t) \rightarrow h(t)$

Recall the definition of approximated impulse function

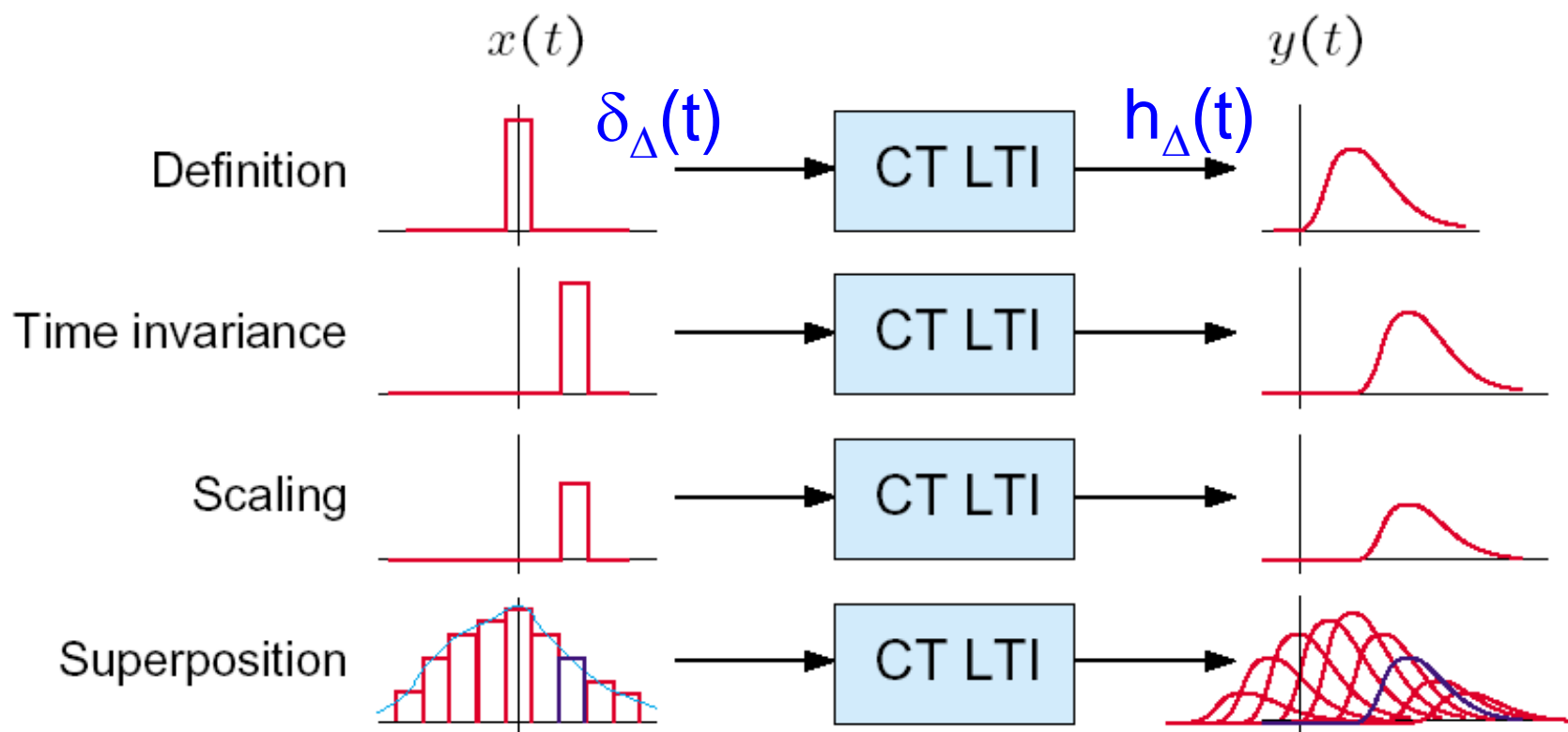


Staircase Approximation of $x(t)$

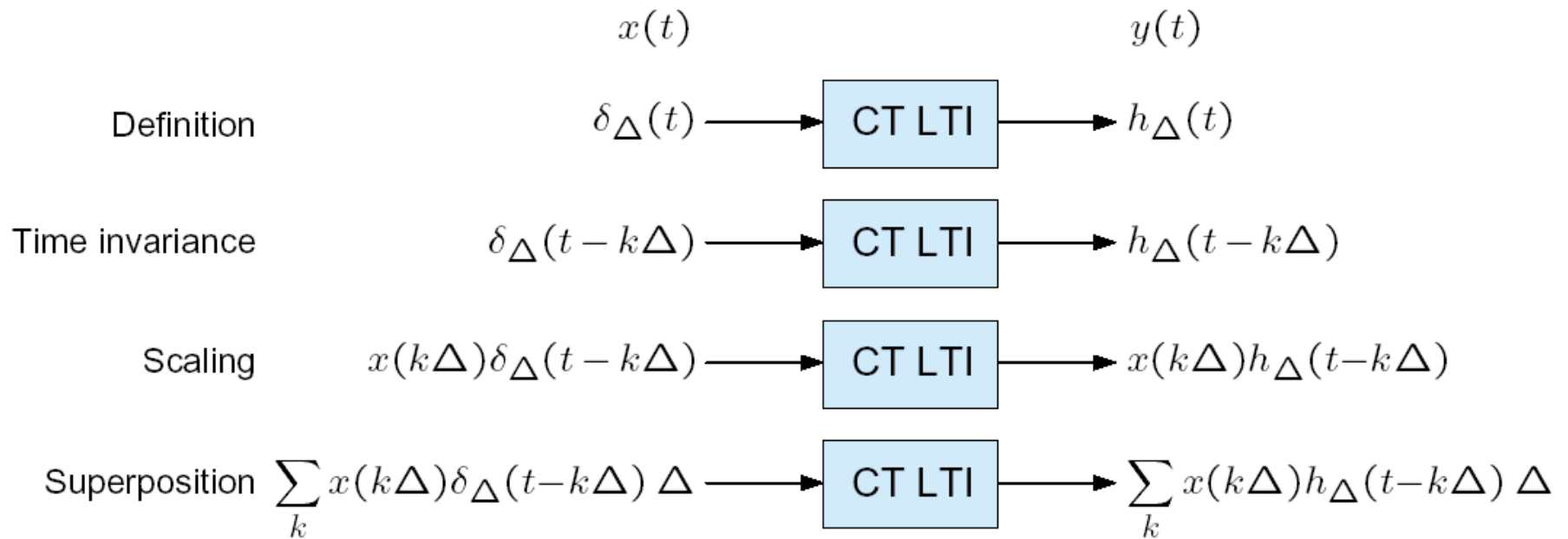
$$x_{\Delta}(t) = \sum_{k=-\infty}^{\infty} x(k\Delta) \delta_{\Delta}(t - k\Delta) \Delta$$



Convolution Integral



Convolution Integral



Convolution Integral

The derivation shows that a staircase approximation to the input

$$x_{\Delta}(t) = \sum_{k=-\infty}^{\infty} x(k\Delta)\delta_{\Delta}(t - k\Delta)\Delta$$

yields an approximation to the output

$$y_{\Delta}(t) = \sum_{k=-\infty}^{\infty} x(k\Delta)h_{\Delta}(t - k\Delta)\Delta$$

Now we take the limit. As $\Delta \rightarrow 0$, $\tau_{\Delta}(t) \rightarrow \delta(t)$, $h_{\Delta}(t) \rightarrow h(t)$, $x_{\Delta}(t) \rightarrow x(t)$, and $y_{\Delta}(t) \rightarrow y(t)$. Also, the sums approach the integrals

$$x(t) = \int_{-\infty}^{\infty} x(\tau)\delta(t - \tau)d\tau \quad \text{Sifting property}$$

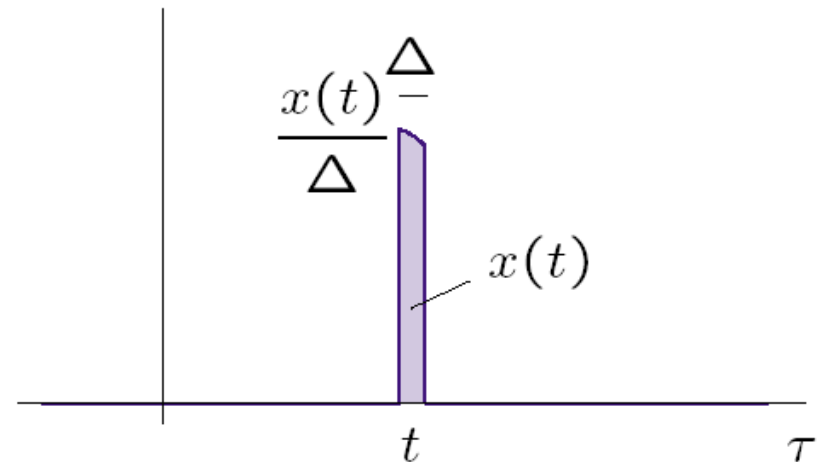
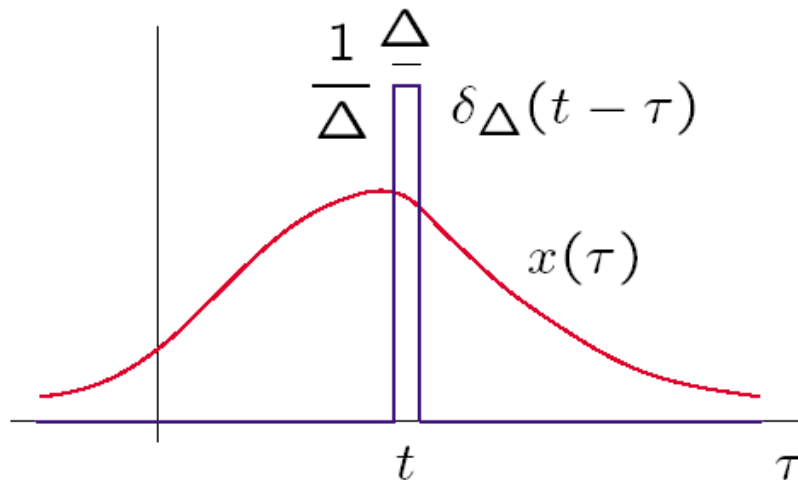
$$y(t) = \int_{-\infty}^{\infty} x(\tau)h(t - \tau)d\tau \quad \text{Convolution integral}$$

Another Interpretation of Sifting Property

To see the meaning of the sifting property

$$x(t) = \int_{-\infty}^{\infty} x(\tau) \delta(t - \tau) d\tau$$

we approximate the impulse with a tall, narrow pulse $\delta_{\Delta}(t - \tau)$



Convolution Integral

- To compute the convolution integral

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

Step 1 Plot x and h vs τ since the convolution integral is on τ .

Step 2 Flip $h(\tau)$ around the vertical axis to obtain $h(-\tau)$.

Step 3 Shift $h(-\tau)$ by t to obtain $h(t - \tau)$.

Step 4 Multiply to obtain $x(\tau) h(t - \tau)$.

Step 5 Integrate on τ to compute $\int x(\tau) h(t - \tau) d\tau$.

Step 6 Increase t and repeat **Steps 3-6**.

Example 1

- Let $x(t)$ be the input to a LTI system with unit impulse response $h(t)$

$$x(t) = e^{-at}u(t) \quad a > 0$$

$$h(t) = u(t)$$

- For $t > 0$

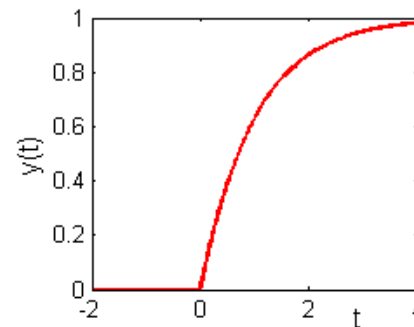
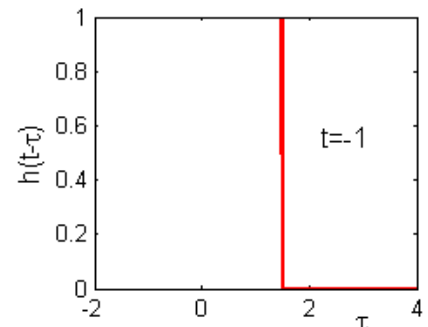
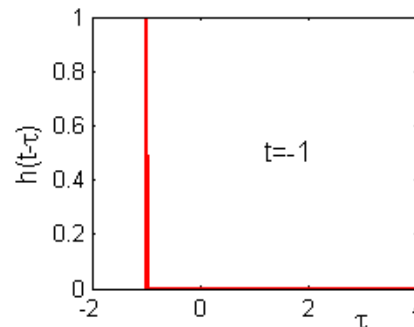
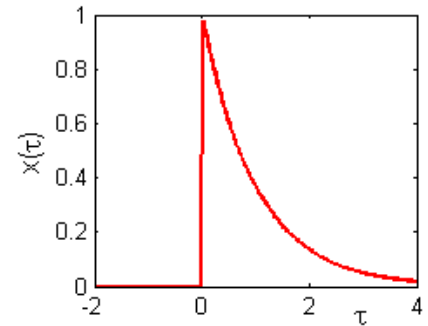
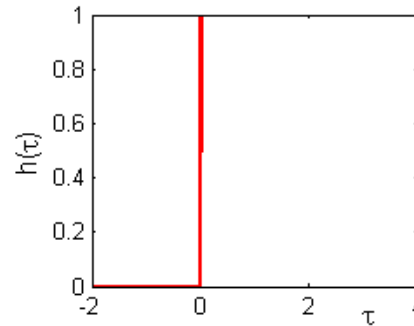
$$x(\tau)h(t-\tau) = \begin{cases} e^{-a\tau} & 0 < \tau < t \\ 0 & \text{otherwise} \end{cases}$$

- We can compute $y(t)$ for $t > 0$

$$\begin{aligned} y(t) &= \int_0^t e^{-a\tau} d\tau = -\frac{1}{a} e^{-a\tau} \Big|_0^t \\ &= \frac{1}{a} (1 - e^{-at}) \end{aligned}$$

- So for all t

$$y(t) = \frac{1}{a} (1 - e^{-at}) u(t)$$



In this example
 $a=1$

Example 2

- Calculate the convolution of the following signals

$$x(t) = e^{2t}u(-t)$$

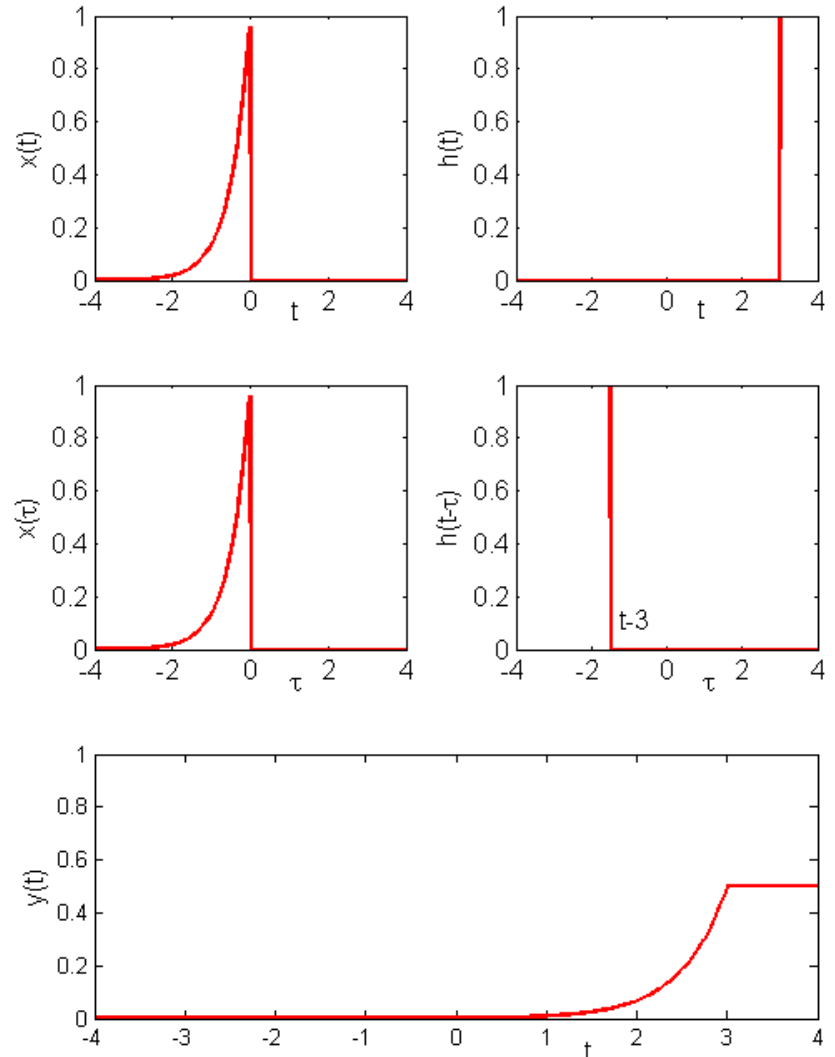
$$h(t) = u(t-3)$$

- For $t < 3$, the convolution integral becomes

$$y(t) = \int_{-\infty}^{t-3} e^{2\tau} d\tau = \frac{1}{2} e^{2(t-3)}$$

- For $t-3 \geq 0$, the product $x(\tau)h(t-\tau)$ is non-zero for $-\infty < \tau < 0$, so the convolution integral becomes

$$y(t) = \int_{-\infty}^0 e^{2\tau} d\tau = \frac{1}{2}$$



Properties of LTI Systems

Properties of Convolution

$$x[n] * y[n] = \sum_{k=-\infty}^{\infty} x[k] y[n-k]$$

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

- Commutative

$$x[n] * y[n] = y[n] * x[n]$$

$$x(t) * y(t) = y(t) * x(t)$$

- Distributive

$$x[n] * (y_1[n] + y_2[n]) = x[n] * y_1[n] + x[n] * y_2[n]$$

$$x(t) * (y_1(t) + y_2(t)) = x(t) * y_1(t) + x(t) * y_2(t)$$

- Associative

$$x[n] * (y_1[n] * y_2[n]) = (x[n] * y_1[n]) * y_2[n]$$

$$x(t) * (y_1(t) * y_2(t)) = (x(t) * y_1(t)) * y_2(t)$$

Causality of LTI Systems

- A system is causal if its output depends only on the past and the present values of the input signal.

- Consider the following for a causal DT LTI system:

$$y[n] = \sum_{k=-\infty}^{\infty} x[k]h[n-k]$$

- ▶ Because of causality $h[n-k]$ must be zero for $k > n$.
 - ▶ In other words, $h[n]=0$ for $n < 0$.
- Similarly for a CT LTI system to be causal
 - ▶ $h(t) = 0$ for $t < 0$.

Causality of LTI Systems

- So the convolution sum for a causal LTI system becomes

$$y[n] = \sum_{k=-\infty}^n x[k]h[n-k] = \sum_{k=0}^{\infty} h[k]x[n-k]$$

- Similarly, the convolution integral for a causal LTI system becomes

$$y[n] = \int_{-\infty}^t x(\tau)h(t-\tau)d\tau = \int_0^{\infty} h(\tau)x(t-\tau)d\tau$$

- So, if a given system is causal, one can infer that its impulse response is zero for negative time values, and use the above simpler convolution formulas.

Stability of LTI Systems

- A system is stable if a bounded input yields a bounded output (BIBO). In other words, if $|x[n]| < k_1$ then $|y[n]| < k_2$.

- Note that

$$|y[n]| = \left| \sum_{k=-\infty}^{\infty} x[n-k]h[k] \right| \leq \sum_{k=-\infty}^{\infty} |x[n-k]| |h[k]| \leq k_1 \sum_{k=-\infty}^{\infty} |h[k]|$$

- Therefore, a DT system is stable if

$$\sum_{k=-\infty}^{\infty} |h[k]| < \infty$$

- Similarly, a CT system is stable if

$$\int_{-\infty}^{\infty} |h(t)| dt < \infty$$