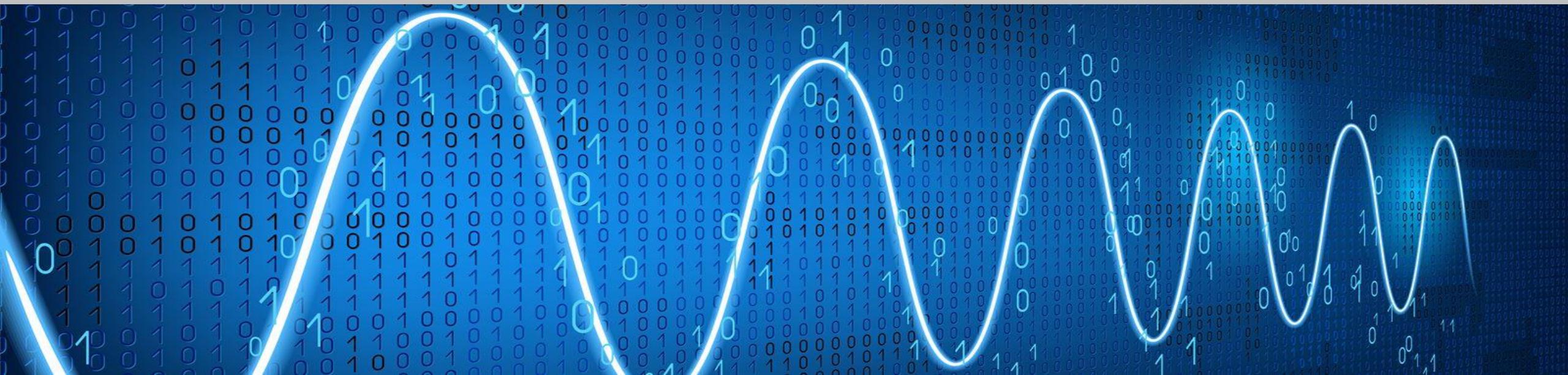


Digital Signal Processing

Lab 05: Analyzing Continuous-Time Systems

Abdallah El Ghamry



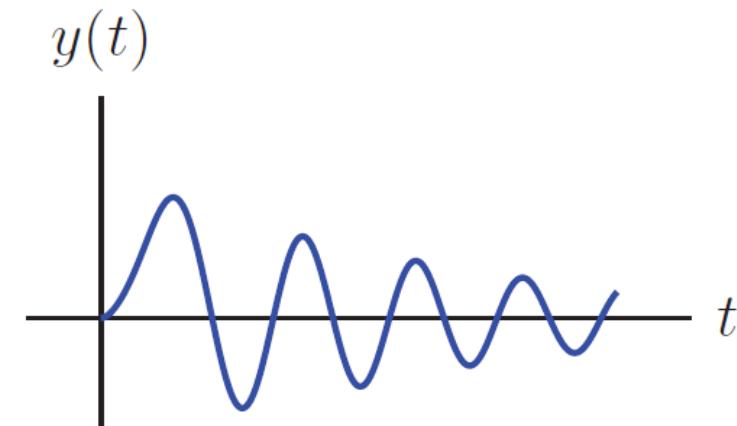
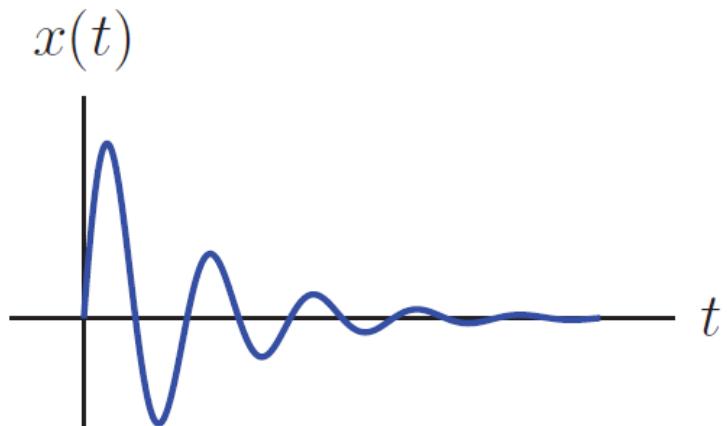
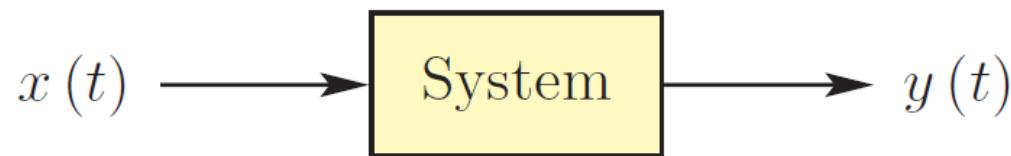
Analyzing Continuous-Time Systems in the Time Domain

The purpose of this lab is to

- Develop the notion of a continuous-time system.
- Discuss the concepts of **linearity** and **time invariance**.
- Learn how to compute the output signal for a linear and time-invariant system using **convolution**.
- Understand the graphical interpretation of the steps involved in carrying out the **convolution operation**.
- Learn the concepts of **causality** and **stability**.

System

- A **system** is any physical entity that takes in a set of one or more physical signals and, in response, produces a new set of one or more physical signals.



System

- The **mathematical model** of a system is a function, formula or algorithm to approximately recreate the same cause-effect relationship between the mathematical models of the input and the output signals.
- The **relationship between the input and the output signals** of a continuous-time system will be mathematically modeled as

$$y(t) = \text{Sys}\{x(t)\}$$

- The operator $\text{Sys}\{\dots\}$ represents the **transformation** applied to $x(t)$.

System

- A system that **amplifies its input signal** by a constant **gain factor K** to yield an output signal

$$y(t) = Kx(t)$$

- A system that **delays its input signal** by a constant **time delay τ** to produce

$$y(t) = x(t - \tau)$$

- A system that produces an output signal that is **proportional to the square of the input signal** as in

$$y(t) = K[x(t)]^2$$

Linearity

- A system is said to be **linear** if the mathematical transformation

$$y(t) = \text{Sys}\{x(t)\}$$

satisfies the following two equations for any two input signals $x_1(t)$, $x_2(t)$ and any arbitrary constant gain factor α_1 .

$$\text{Sys}\{x_1(t) + x_2(t)\} = \text{Sys}\{x_1(t)\} + \text{Sys}\{x_2(t)\}$$

$$\text{Sys}\{\alpha_1 x_1(t)\} = \alpha_1 \text{Sys}\{x_1(t)\}$$

Linearity

- The **additivity rule** can be stated as follows:

The response of a linear system to **the sum of two signals** is **the same as the sum of individual responses** to each of the two input signals.

$$\text{Sys}\{x_1(t) + x_2(t)\} = \text{Sys}\{x_1(t)\} + \text{Sys}\{x_2(t)\}$$

- The **homogeneity rule** can be stated as follows:

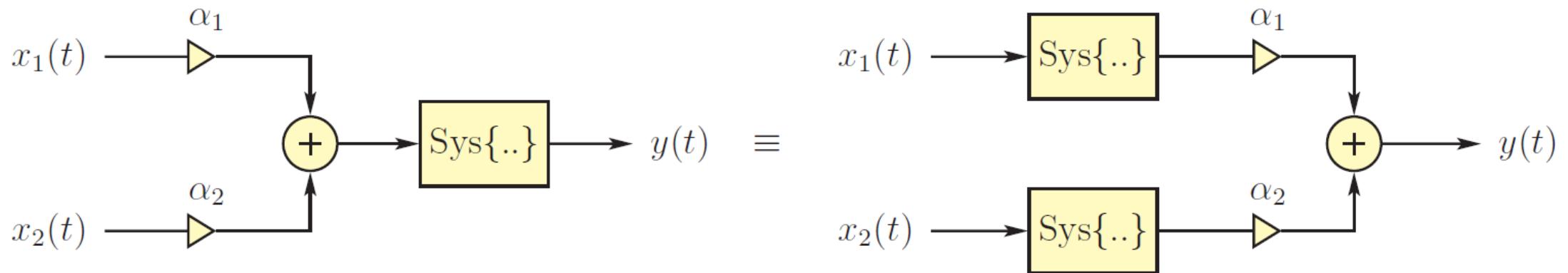
Scaling the input signal of a linear system by a constant gain factor causes the output signal to be **scaled with the same gain factor**.

$$\text{Sys}\{\alpha_1 x_1(t)\} = \alpha_1 \text{Sys}\{x_1(t)\}$$

Linearity: Superposition Principle

- The two criteria can be combined into one equation which is referred to as the **superposition principle**.

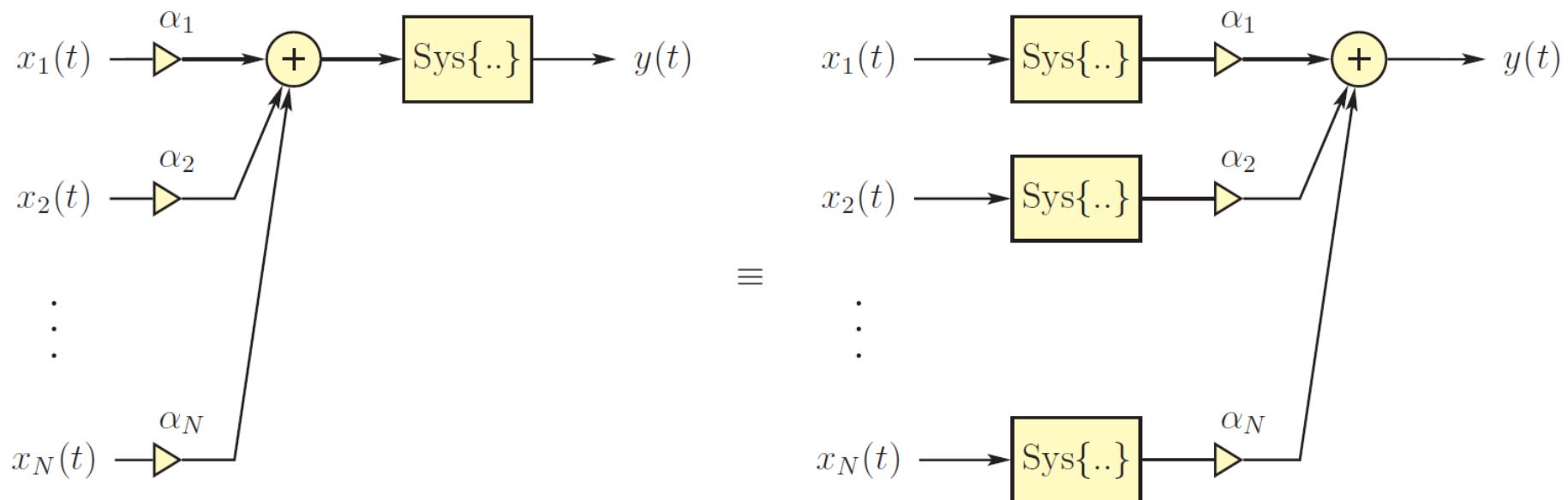
$$\text{Sys}\{\alpha_1 x_1(t) + \alpha_2 x_2(t)\} = \alpha_1 \text{Sys}\{x_1(t)\} + \alpha_2 \text{Sys}\{x_2(t)\}$$



Linearity: Superposition Principle

- The generalized form of the superposition principle can be expressed verbally as follows:

$$y(t) = \text{Sys} \left\{ \sum_{i=1}^N \alpha_i x_i(t) \right\} = \sum_{i=1}^N \alpha_i y_i(t)$$



Example 2.1

Four different systems are described below through their input-output relationships.

For each, determine if the system is linear or not:

- a. $y(t) = 5x(t)$
- b. $y(t) = 5x(t) + 3$
- c. $y(t) = 3 [x(t)]^2$
- d. $y(t) = \cos(x(t))$

Example 2.1 (a) – Solution

- a. If two input signals $x_1(t)$ and $x_2(t)$ are applied to the system individually, they produce the output signals $y_1(t) = 5x_1(t)$ and $y_2(t) = 5x_2(t)$ respectively. Let the input signal be $x(t) = \alpha_1 x_1(t) + \alpha_2 x_2(t)$. The corresponding output signal is found using the system definition:

$$\begin{aligned}y(t) &= 5x(t) \\&= 5[\alpha_1 x_1(t) + \alpha_2 x_2(t)] \\&= \alpha_1 [5x_1(t)] + \alpha_2 [5x_2(t)] \\&= \alpha_1 y_1(t) + \alpha_2 y_2(t)\end{aligned}$$

Superposition principle holds; therefore this system is linear.

Example 2.1 (b) – Solution

- b. If two input signals $x_1(t)$ and $x_2(t)$ are applied to the system individually, they produce the output signals $y_1(t) = 5x_1(t) + 3$ and $y_2(t) = 5x_2(t) + 3$ respectively.

We will again use the combined input signal $x(t) = \alpha_1 x_1(t) + \alpha_2 x_2(t)$ for testing. The corresponding output signal for this system is

$$\begin{aligned}y(t) &= 5x(t) + 3 \\&= 5\alpha_1 x_1(t) + 5\alpha_2 x_2(t) + 3\end{aligned}$$

The output signal $y(t)$ cannot be expressed in the form $y(t) = \alpha_1 y_1(t) + \alpha_2 y_2(t)$.

Superposition principle does **not hold true** in this case.

The system is **not linear**.

Example 2.1 (c) – Solution

- c. Using two input signals $x_1(t)$ and $x_2(t)$ individually, the corresponding output signals produced by this system are $y_1(t) = 3 [x_1(t)]^2$ and $y_2(t) = 3 [x_2(t)]^2$ respectively. Applying the linear combination $x(t) = \alpha_1 x_1(t) + \alpha_2 x_2(t)$ to the system produces the output signal

$$\begin{aligned}y(t) &= 3 [\alpha_1 x_1(t) + \alpha_2 x_2(t)]^2 \\&= 3 \alpha_1^2 [x_1(t)]^2 + 6 \alpha_1 \alpha_2 x_1(t) x_2(t) + 3 \alpha_2^2 [x_2(t)]^2\end{aligned}$$

Superposition principle does **not hold true** in this case.

The system is **not linear**.

Example 2.1 (d) – Solution

- d. The test signals $x_1(t)$ and $x_2(t)$ applied to the system individually produce the output signals $y_1(t) = \cos[x_1(t)]$ and $y_2(t) = \cos[x_2(t)]$ respectively. Their linear combination $x(t) = \alpha_1 x_1(t) + \alpha_2 x_2(t)$ produces the output signal

$$y(t) = \cos[\alpha_1 x_1(t) + \alpha_2 x_2(t)]$$

Superposition principle does **not hold true** in this case.

The system is **not linear**.

Time Invariance

- A system is said to be **time-invariant** if its **behavior characteristics** do not change in time.
- Consider a **continuous-time system** with the input-output relationship

$$\text{Sys}\{x(t)\} = y(t)$$

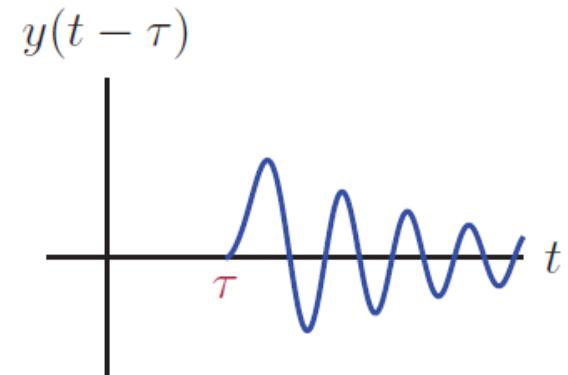
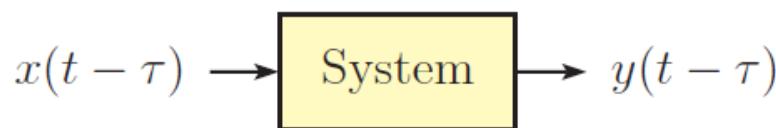
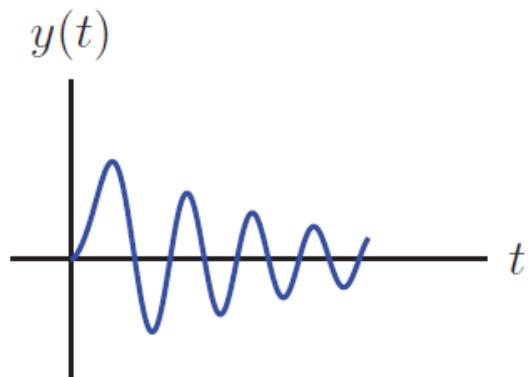
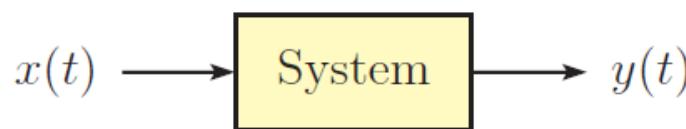
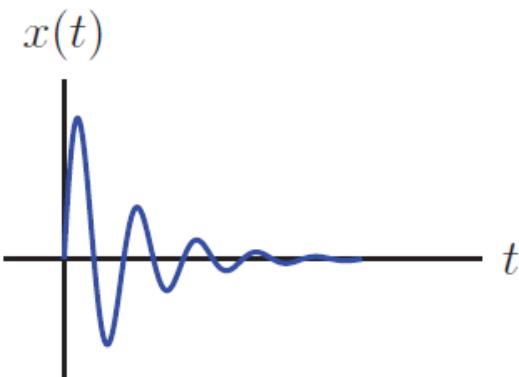
- If the input signal applied to a time-invariant system is **time-shifted by τ seconds**, the only effect of this delay should be to cause an **equal amount of time shift in the output signal**, but to otherwise **leave the shape of the output signal unchanged**.

$$\text{Sys}\{x(t - \tau)\} = y(t - \tau)$$

Time Invariance

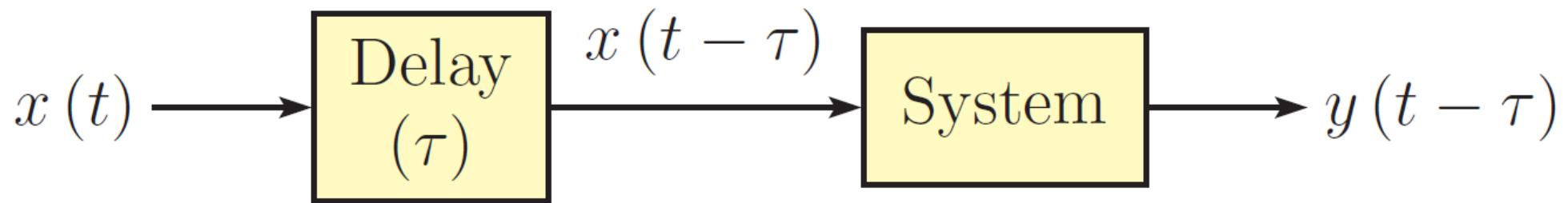
- Condition for **time-invariance**:

$$\text{Sys}\{x(t)\} = y(t) \quad \text{implies that} \quad \text{Sys}\{x(t - \tau)\} = y(t - \tau)$$

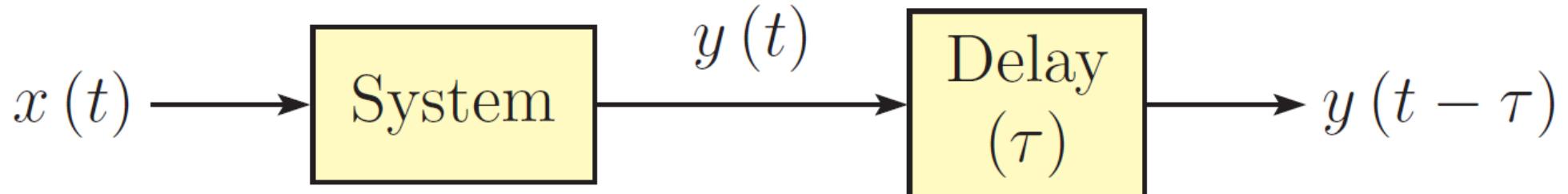


Time Invariance

- Alternatively, the relationship described can be characterized by the equivalence of the two system configurations shown in Figure.



(a)



Example 2.2

Three different systems are described below through their input-output relationships.

For each, determine whether the system is time-invariant or not:

a. $y(t) = 5x(t)$

b. $y(t) = 3 \cos(x(t))$

c. $y(t) = 3 \cos(t)x(t)$

Example 2.2 (a) – Solution

- a. For this system, if the input signal $x(t)$ is delayed by τ seconds, the corresponding output signal would be

$$\text{Sys}\{x(t - \tau)\} = 5x(t - \tau) = y(t - \tau)$$

This system is time-invariant.

Example 2.2 (b) – Solution

b. Let the input signal be $x(t - \tau)$. The output of the system is

$$\text{Sys}\{x(t - \tau)\} = 3 \cos(x(t - \tau)) = y(t - \tau)$$

This system is **time-invariant**.

Example 2.2 (c) – Solution

c. Again using the delayed input signal $x(t - \tau)$ we obtain the output

$$\text{Sys}\{x(t - \tau)\} = 3 \cos(t) x(t - \tau) \neq y(t - \tau)$$

In this case the system is **not time-invariant** since the time-shifted input signal leads to a response that is not the same as a similarly time-shifted version of the original output signal.

Problem 2.1

A number of systems are specified below in terms of their input-output relationships.

For each case, determine if the system is linear and/or time-invariant.

a. $y(t) = |x(t)| + x(t)$

b. $y(t) = t x(t)$

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

Problem 2.1 (a) – Solution

a. $y(t) = |x(t)| + x(t)$

$$y_1(t) = \text{Sys}\{x_1(t)\} = |x_1(t)| + x_1(t)$$

$$y_2(t) = \text{Sys}\{x_2(t)\} = |x_2(t)| + x_2(t)$$

Using $x(t) = \alpha_1 x_1(t) + \alpha_2 x_2(t)$ as input we obtain

$$\begin{aligned} y(t) &= \text{Sys}\{\alpha_1 x_1(t) + \alpha_2 x_2(t)\} \\ &= |\alpha_1 x_1(t) + \alpha_2 x_2(t)| + \alpha_1 x_1(t) + \alpha_2 x_2(t) \\ &\neq \alpha_1 y_1(t) + \alpha_2 y_2(t) \end{aligned}$$

The system is not linear.

$$\text{Sys}\{x_1(t-\tau)\} = |x_1(t-\tau)| + x_1(t-\tau) = y_1(t-\tau)$$

The system is time-invariant.

Problem 2.1 (b) – Solution

b. $y(t) = t x(t)$

$$y_1(t) = \text{Sys}\{x_1(t)\} = t x_1(t)$$

$$y_2(t) = \text{Sys}\{x_2(t)\} = t x_2(t)$$

Using $x(t) = \alpha_1 x_1(t) + \alpha_2 x_2(t)$ as input we obtain

$$\begin{aligned} y(t) &= \text{Sys}\{\alpha_1 x_1(t) + \alpha_2 x_2(t)\} \\ &= t [\alpha_1 x_1(t) + \alpha_2 x_2(t)] \\ &= \alpha_1 t x_1(t) + \alpha_2 t x_2(t) \\ &= \alpha_1 y_1(t) + \alpha_2 y_2(t) \end{aligned}$$

The system is linear.

$$\text{Sys}\{x_1(t - \tau)\} = t x_1(t - \tau) \neq y_1(t - \tau)$$

The system is not time-invariant.

Problem 2.1 (c) – Solution

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

$$y_1(t) = \text{Sys}\{x_1(t)\} = e^{-t} x_1(t)$$

$$y_2(t) = \text{Sys}\{x_2(t)\} = e^{-t} x_2(t)$$

Using $x(t) = \alpha_1 x_1(t) + \alpha_2 x_2(t)$ as input we obtain

$$\begin{aligned} y(t) &= \text{Sys}\{\alpha_1 x_1(t) + \alpha_2 x_2(t)\} \\ &= e^{-t} [\alpha_1 x_1(t) + \alpha_2 x_2(t)] \\ &= \alpha_1 e^{-t} x_1(t) + \alpha_2 e^{-t} x_2(t) \\ &= \alpha_1 y_1(t) + \alpha_2 y_2(t) \end{aligned}$$

The system is linear.

$$\text{Sys}\{x_1(t - \tau)\} = e^{-t} x_1(t - \tau) \neq y_1(t - \tau)$$

The system is not time-invariant.

Problem 2.2

2.2. Consider the cascade combination of two systems shown in Fig. P.2.2(a).



Figure P. 2.2

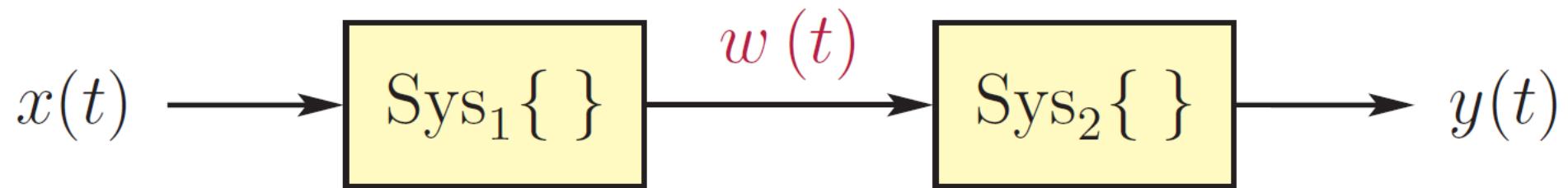
a. Let the input-output relationships of the two subsystems be given as

$$\text{Sys}_1\{x(t)\} = 3x(t) \quad \text{and} \quad \text{Sys}_2\{w(t)\} = w(t - 2)$$

Write the relationship between $x(t)$ and $y(t)$.

b. Let the order of the two subsystems be changed as shown in Fig. P.2.2(b). Write the relationship between $x(t)$ and $\bar{y}(t)$. Does changing the order of two subsystems change the overall input-output relationship of the system?

Problem 2.2 (a) – Solution

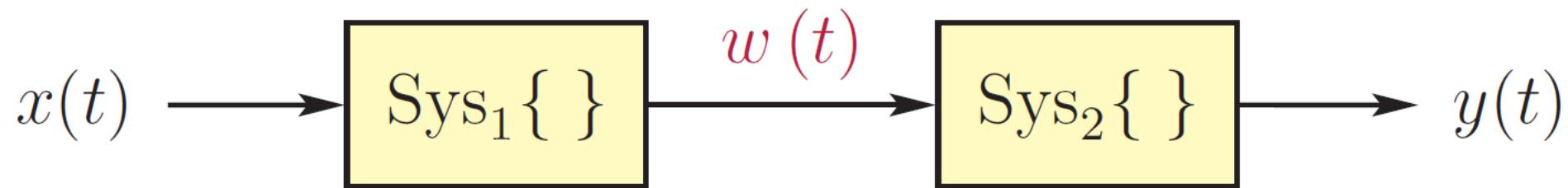


$$\text{Sys}_1 \{x(t)\} = 3x(t) \quad \text{and} \quad \text{Sys}_2 \{w(t)\} = w(t - 2)$$

$$w(t) = 3x(t)$$

$$y(t) = w(t - 2) = 3x(t - 2)$$

Problem 2.2 (b) – Solution



$$\text{Sys}_1 \{x(t)\} = 3x(t) \quad \text{and} \quad \text{Sys}_2 \{w(t)\} = w(t - 2)$$

$$\bar{w}(t) = x(t - 2)$$

$$\bar{y}(t) = 3\bar{w}(t) = 3x(t - 2)$$

Input-output relationship of the system **does not change** when the order of the two subsystems is changed.

Problem 2.3 (b)

2.3. Repeat Problem 2.2 with the following sets of subsystems:

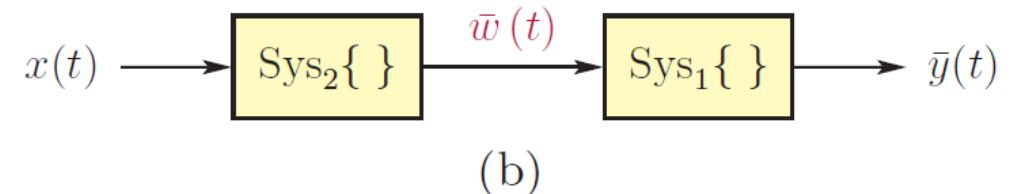
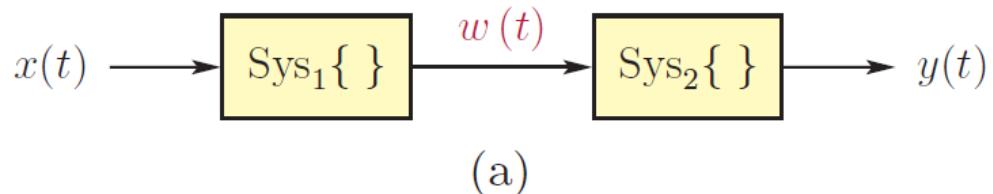


Figure P. 2.2

- a. Let the input-output relationships of the two subsystems be given as

$$\text{Sys}_1 \{x(t)\} = 3x(t) \quad \text{and} \quad \text{Sys}_2 \{w(t)\} = w(t) + 5$$

Write the relationship between $x(t)$ and $y(t)$.

- b. Let the order of the two subsystems be changed as shown in Fig. P.2.2(b). Write the relationship between $x(t)$ and $\bar{y}(t)$. Does changing the order of two subsystems change the overall input-output relationship of the system?

Problem 2.3 (b) – Solution

Using the first configuration:

$$w(t) = 3x(t)$$

$$y(t) = w(t) + 5 = 3x(t) + 5$$

Using the second configuration:

$$\bar{w}(t) = x(t) + 5$$

$$\bar{y}(t) = 3\bar{w}(t) = 3[x(t) + 5] = 3x(t) + 15$$

Input-output relationship of the system **changes** when the order of the two subsystems is changed.

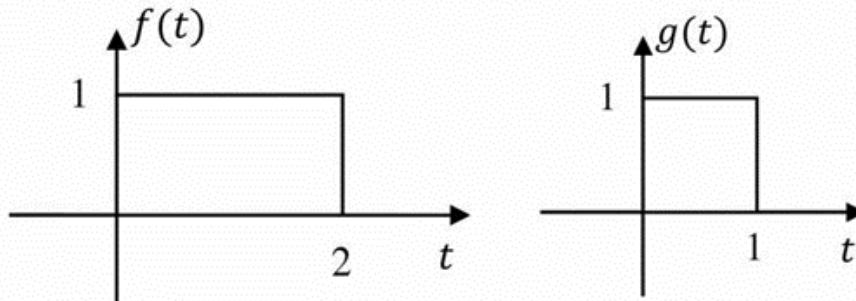
- We will work with **continuous-time systems** that are both **linear** and **time-invariant**.
- A number of **time-** and **frequency-domain analysis** and **design** techniques will be developed for such systems.
- For simplicity, we will use the acronym **CTLTI** to refer to **continuous-time linear and time-invariant systems**.

Convolution

- A **convolution** is an integral that expresses the amount of overlap of one function when it is **shifted** over another function.

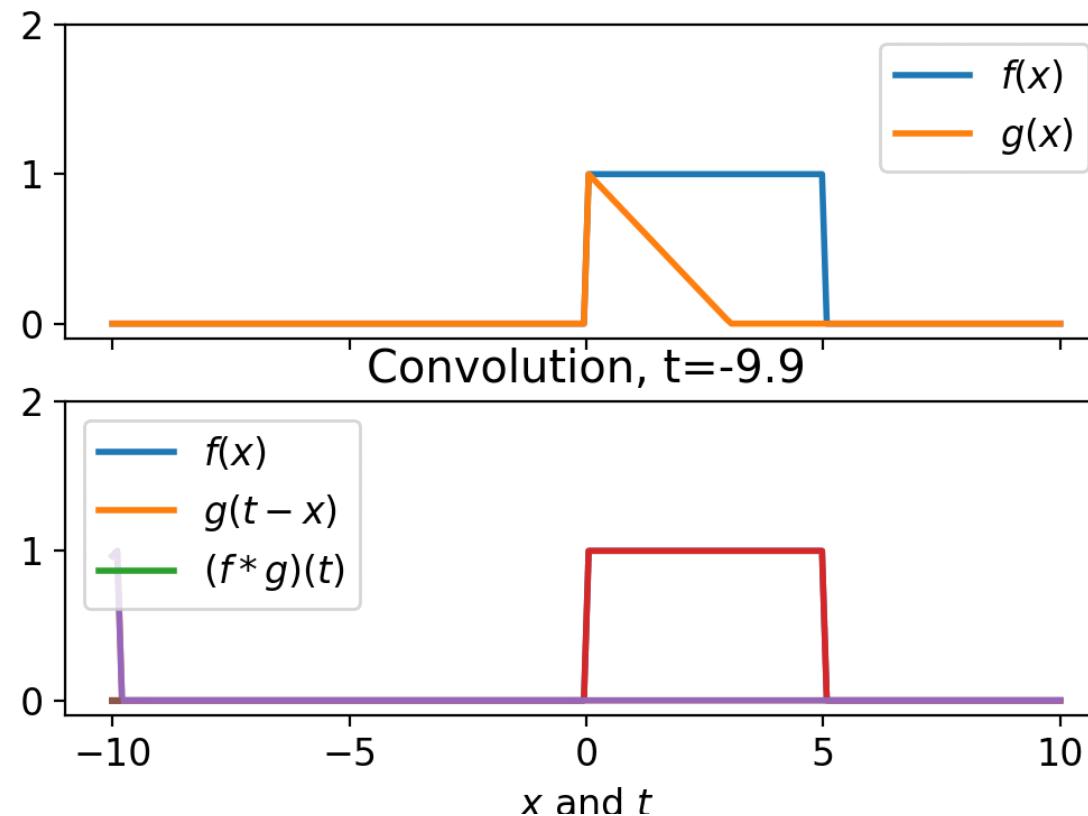
Convolution of two box functions

- $f(t) * g(t)$



Convolution

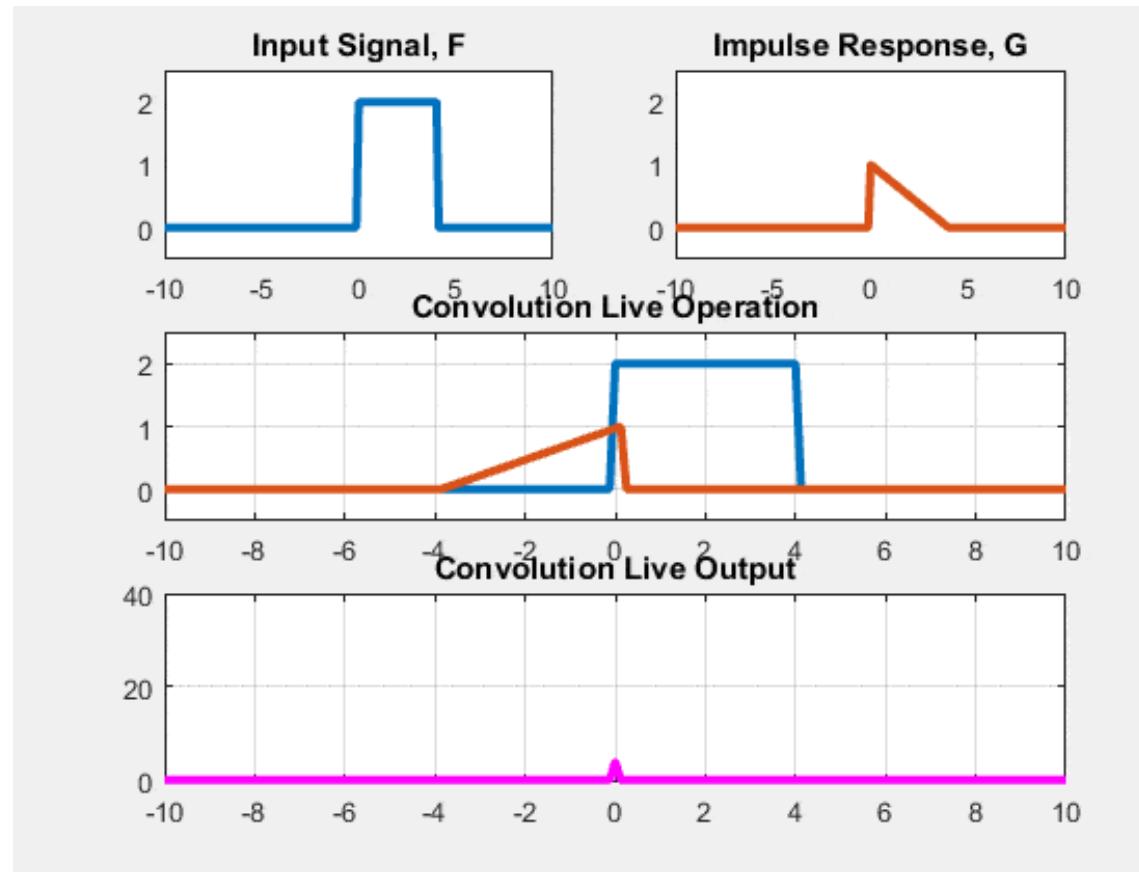
- A convolution is an integral that expresses the amount of overlap of one function when it is shifted over another function.



https://mahfuzdotsite.files.wordpress.com/2018/12/convolution_anim1.gif

Convolution

- A **convolution** is an integral that expresses the amount of overlap of one function when it is **shifted** over another function.



http://quincyaf Flint.weebly.com/uploads/2/6/5/0/26500868/convolution-1-fast_orig.gif

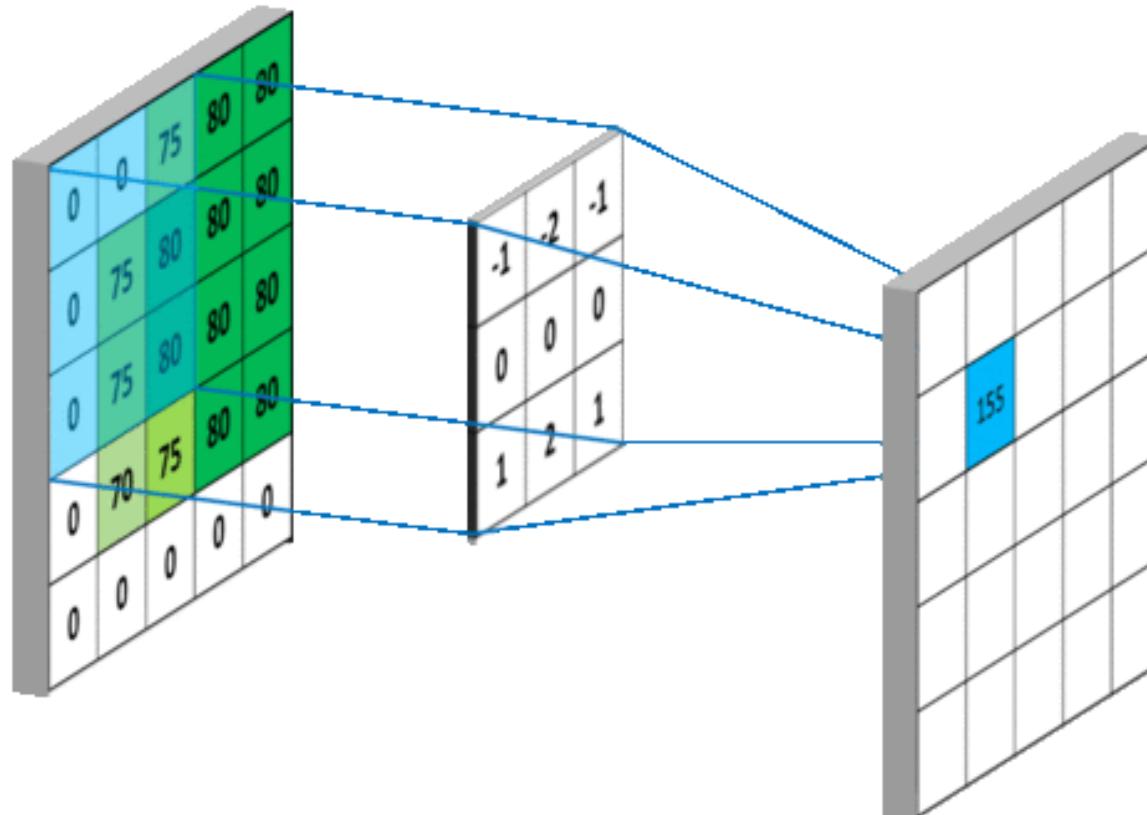
Convolution

- A **convolution** is an integral that expresses the amount of overlap of one function when it is **shifted** over another function.

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

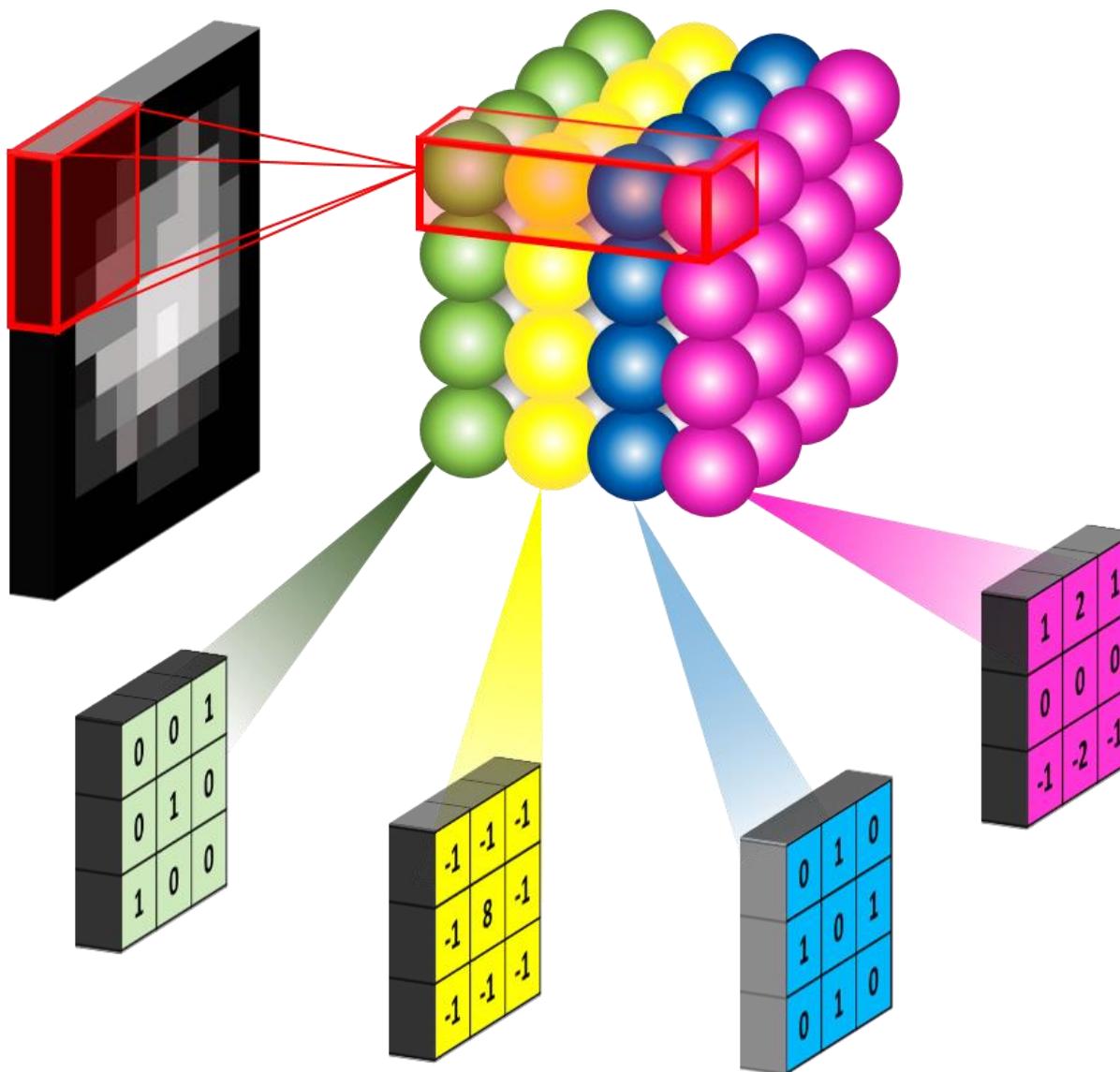
$$y(t) = h(t) * x(t) = \int_{-\infty}^{\infty} h(\lambda) x(t - \lambda) d\lambda$$

Convolutional Neural Networks

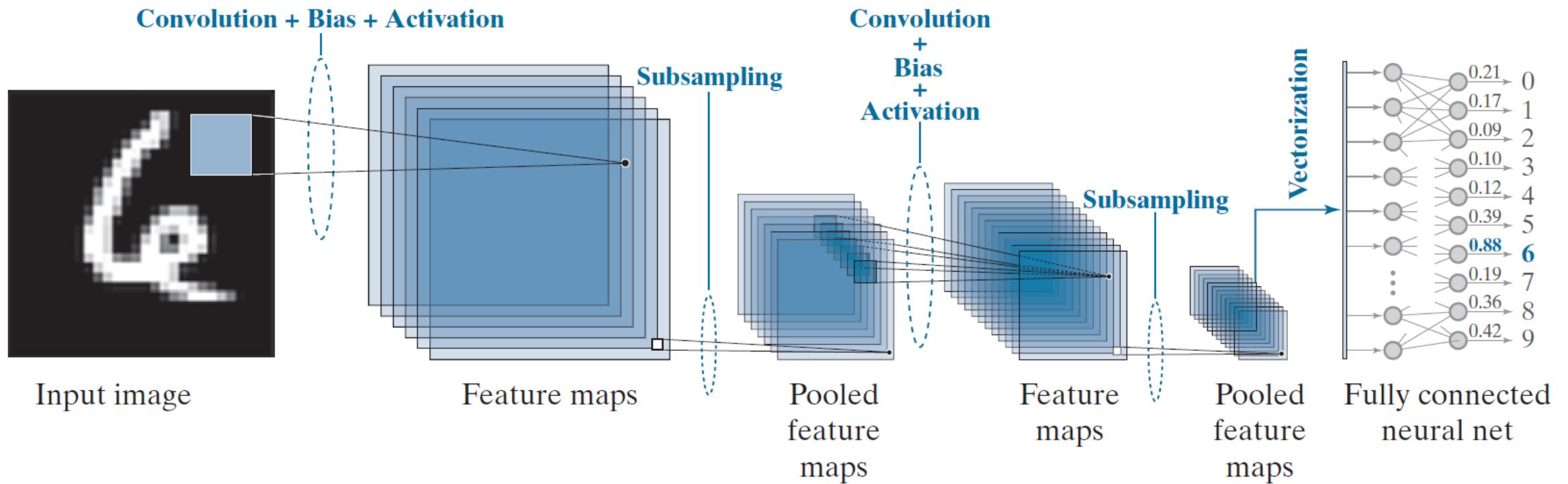


<https://mlnotebook.github.io/post/CNN1/>

Convolutional Neural Networks



Convolutional Neural Networks



Convolution Operation for CTLTI Systems

- The **output signal** $y(t)$ of a CTLTI system is obtained by **convolving** the **input signal** $x(t)$ and the **impulse response** $h(t)$ of the system.
- This relationship is expressed in compact notation as

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

where the symbol $*$ represents the **convolution operator**.

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

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

Unit-Impulse Function

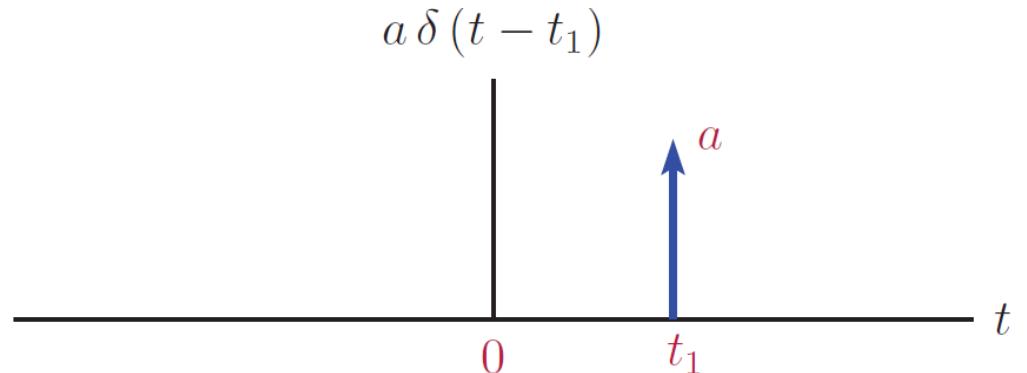
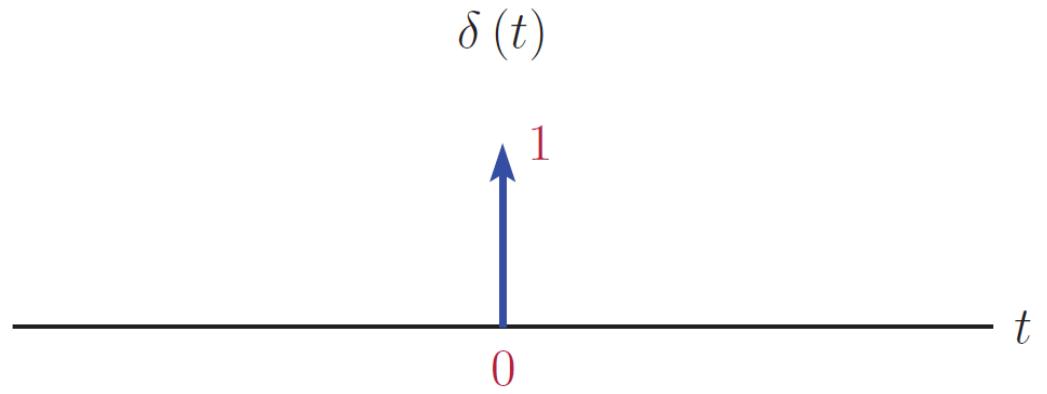
- The unit-impulse function plays an important role in mathematical modeling and analysis of signals and linear systems.

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

$$\int_{-\infty}^{\infty} \delta(t) dt = 1$$

$$a \delta(t - t_1) = \begin{cases} 0, & \text{if } t \neq t_1 \\ \text{undefined}, & \text{if } t = t_1 \end{cases}$$

$$\int_{-\infty}^{\infty} a \delta(t - t_1) dt = a$$



Unit-Impulse Function: Shifting Property

- The **shifting property** of the unit-impulse function is given by

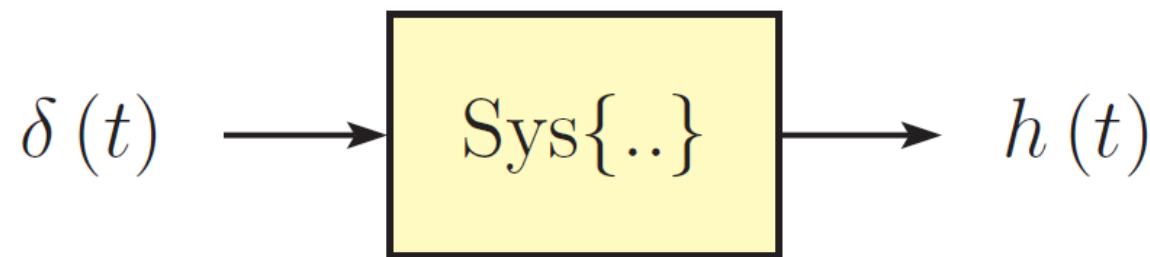
$$x(t) = \int_{-\infty}^{\infty} x(\lambda) \delta(t - \lambda) d\lambda$$

$$x(\lambda) \delta(t - \lambda) = 0, \quad t \neq \lambda$$

$$\begin{aligned} \int_{-\infty}^{\infty} x(\lambda) \delta(t - \lambda) d\lambda &= \int_{-\infty}^{\infty} x(t) \delta(t - \lambda) d\lambda \\ &= x(t) \int_{-\infty}^{\infty} \delta(t - \lambda) d\lambda \\ &= x(t) \end{aligned}$$

Impulse Response

- In previous sections, we have explored that a CTLTI system can be described by constant-coefficient **ordinary differential equation**.
- An **alternative description** of a CTLTI system can be given in terms of its **impulse response** $h(t)$ which is simply the forced response of the system under consideration **when the input signal is a unit impulse**.



Problem 2.22

2.22. Using the convolution integral given by Eqns. (2.153) and (2.154) prove each of the relationships below:

a. $x(t) * \delta(t) = x(t)$

b. $x(t) * \delta(t - t_0) = x(t - t_0)$

Problem 2.22 – Solution

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

b. $x(t) * \delta(t - t_0) = \int_{-\infty}^{\infty} x(\tau) \delta(t - t_0 - \tau) d\tau = x(t - t_0)$

Problem 2.23

2.23. The impulse response of a CTLTI system is

$$h(t) = \delta(t) - \delta(t - 1)$$

Determine sketch the response of this system to the triangular waveform shown in Fig. P.2.23.

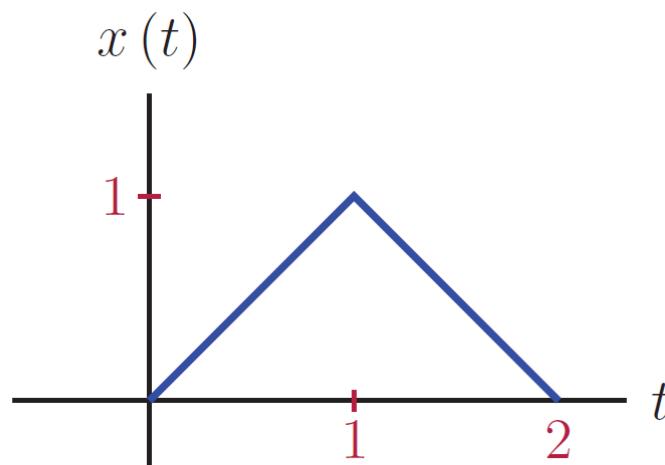


Figure P. 2.23

Problem 2.23 – Solution

$$h(t) = \delta(t) - \delta(t-1)$$

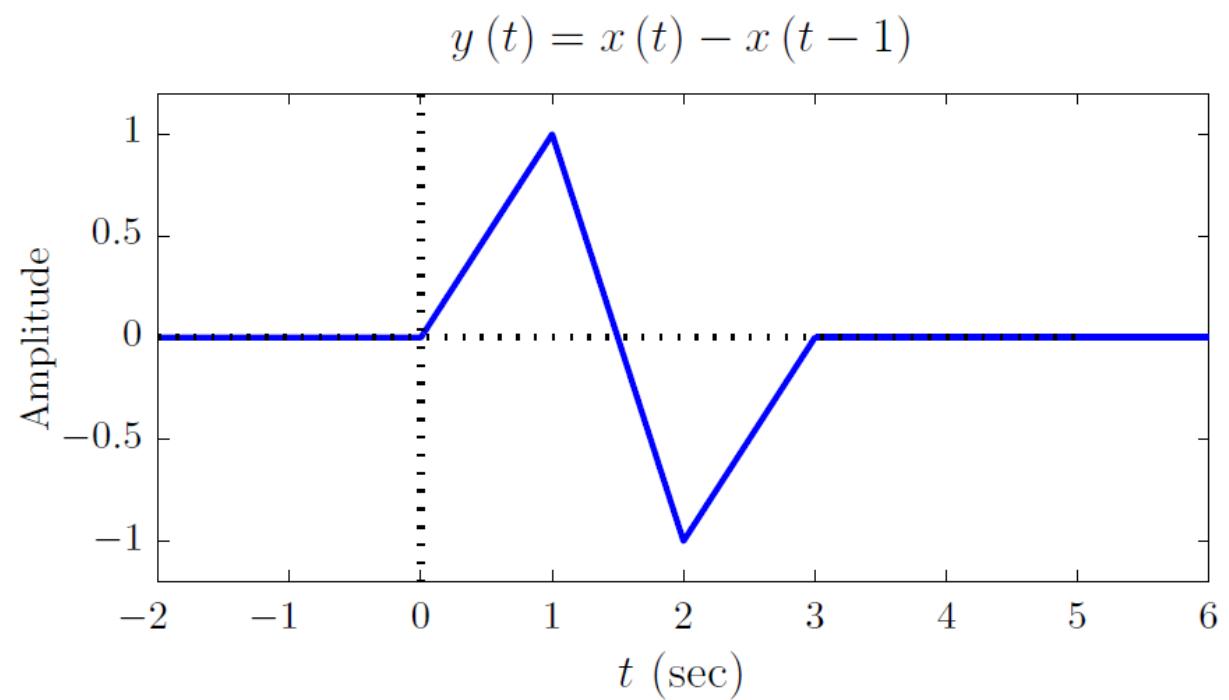
$$\begin{aligned}y(t) &= x(t) * h(t) = \int_{-\infty}^{\infty} x(\lambda) h(t - \lambda) d\lambda \\&= \int_{-\infty}^{\infty} x(\lambda) \delta(t - \lambda) - \delta(t - 1 - \lambda) d\lambda \\&= \int_{-\infty}^{\infty} x(\lambda) \delta(t - \lambda) d\lambda - \int_{-\infty}^{\infty} x(\lambda) \delta(t - 1 - \lambda) d\lambda \\&= x(t) \int_{-\infty}^{\infty} \delta(t - \lambda) d\lambda - x(t-1) \int_{-\infty}^{\infty} \delta(t - 1 - \lambda) d\lambda \\&= x(t) - x(t-1)\end{aligned}$$

Problem 2.23 – Solution

$$x(t) = \begin{cases} t, & 0 < t < 1 \\ -t + 2, & 1 < t < 2 \\ 0, & \text{otherwise} \end{cases}$$

$$x(t-1) = \begin{cases} t-1, & 1 < t < 2 \\ -t+3, & 2 < t < 3 \\ 0, & \text{otherwise} \end{cases}$$

$$y(t) = x(t) - x(t-1) = \begin{cases} t, & 0 < t < 1 \\ -2t+3, & 1 < t < 2 \\ t-3, & 2 < t < 3 \\ 0, & \text{otherwise} \end{cases}$$



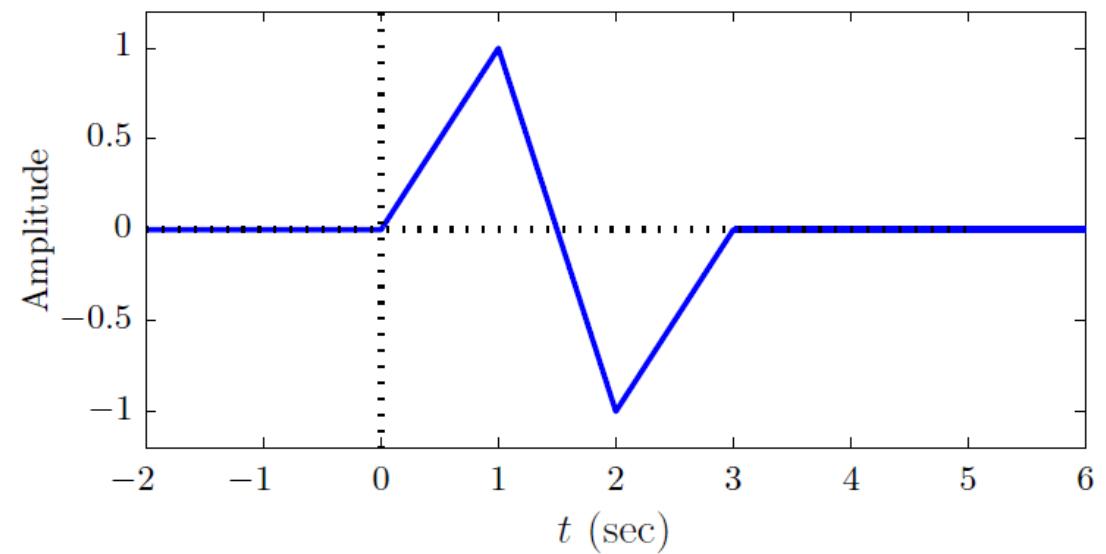
Problem 2.23 – Another Solution

$$\begin{aligned}y(t) &= \int_{-\infty}^{\infty} h(\tau) x(t - \tau) d\tau \\&= \int_{-\infty}^{\infty} [\delta(\tau) - \delta(\tau - 1)] x(t - \tau) d\tau \\&= \int_{-\infty}^{\infty} \delta(\tau) x(t - \tau) d\tau - \int_{-\infty}^{\infty} \delta(\tau - 1) x(t - \tau) d\tau\end{aligned}$$

Using the sifting property of the unit-impulse function, we have

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

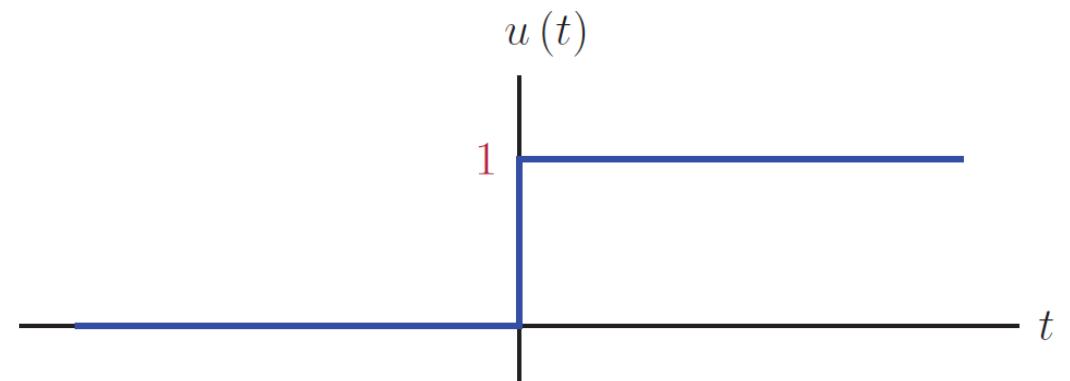
$$y(t) = x(t) - x(t - 1)$$



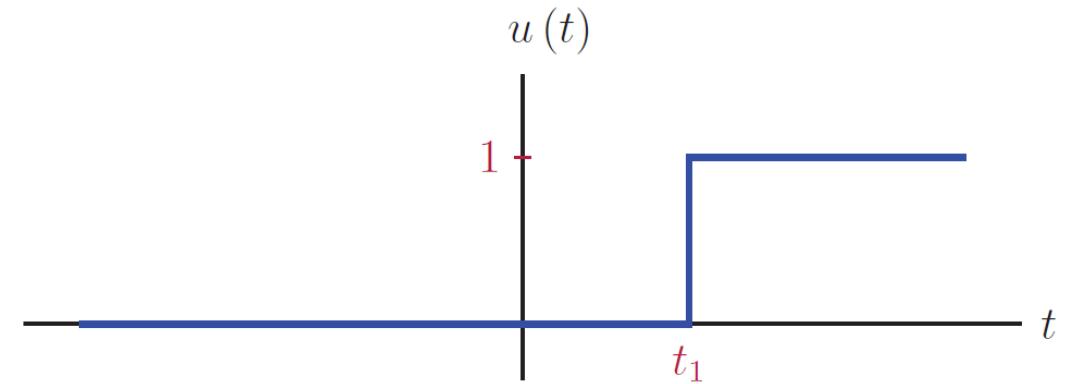
Unit-Step Function

- The unit-step function is useful in situations where we need to model a signal that is turned on or off at a specific time instant.

$$u(t) = \begin{cases} 1, & t > 0 \\ 0, & t < 0 \end{cases}$$



$$u(t - t_1) = \begin{cases} 1, & t > t_1 \\ 0, & t < t_1 \end{cases}$$



Problem 2.26

2.26. For each pair of signals $x(t)$ and $h(t)$ given below, find the convolution $y(t) = x(t)*h(t)$. In each case sketch the signals involved in the convolution integral and determine proper integration limits.

a. $x(t) = u(t)$, $h(t) = e^{-2t} u(t)$

c. $x(t) = u(t - 2)$, $h(t) = e^{-2t} u(\bar{t})$

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

Problem 2.26 (a) – Solution

a. $x(t) = u(t)$, $h(t) = e^{-2t} u(t)$

a.

$$y(t) = \int_{-\infty}^{\infty} u(\lambda) e^{-2(t-\lambda)} u(t-\lambda) d\lambda = \int_0^{\infty} e^{-2(t-\lambda)} u(t-\lambda) d\lambda$$

Case 1: $t < 0$

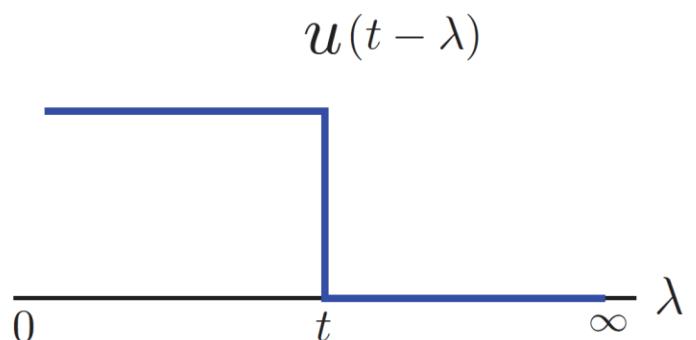
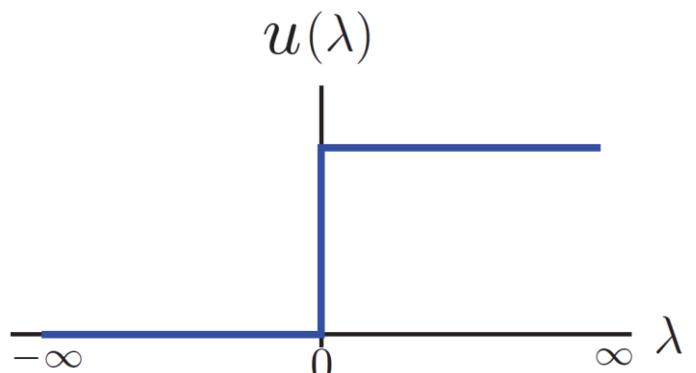
$$y(t) = 0$$

Case 2: $t \geq 0$

$$\int_0^t e^{-2(t-\lambda)} d\lambda = \frac{1}{2} (1 - e^{-2t})$$

Problem 2.26 (a) – Explanation

$$\begin{aligned}y(t) &= x(t) * h(t) = \int_{-\infty}^{\infty} u(\lambda) e^{-2(t-\lambda)} u(t-\lambda) d\lambda \\&= \int_0^{\infty} e^{-2(t-\lambda)} u(t-\lambda) d\lambda \\&= \int_0^t e^{-2(t-\lambda)} d\lambda \\&= \frac{1}{2} e^{-2(t-\lambda)} \Big|_0^t = \frac{1}{2} (1 - e^{-2t})\end{aligned}$$



Problem 2.26 (c) – Solution

c. $x(t) = u(t - 2)$, $h(t) = e^{-2t} u(\bar{t})$

c.

$$y(t) = \int_{-\infty}^{\infty} u(\lambda - 2) e^{-2(t-\lambda)} u(t - \lambda) d\lambda = \int_2^{\infty} e^{-2(t-\lambda)} u(t - \lambda) d\lambda$$

Case 1: $t < 2$

$$y(t) = 0$$

Case 2: $t \geq 2$

$$\int_2^t e^{-2(t-\lambda)} d\lambda = \frac{1}{2} (1 - e^{-2(t-2)})$$

Problem 2.26 (e) – Solution

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

e.

$$y(t) = \int_{-\infty}^{\infty} e^{-\lambda} u(\lambda) e^{-2(t-\lambda)} u(t-\lambda) d\lambda = \int_0^{\infty} e^{-2t+\lambda} u(t-\lambda) d\lambda$$

Case 1: $t < 0$

$$y(t) = 0$$

Case 2: $t \geq 0$

$$y(t) = \int_0^t e^{-2t+\lambda} d\lambda = e^{-t} - e^{-2t}$$

Causality in Continuous-Time Systems

- A system is said to be **causal** if the current value of the output signal depends only on current and past values of the input signal, but **not on its future values**.
- The system with input-output relationship is **causal** since the output signal can be computed based on current and past values of the input signal.

$$y(t) = x(t) + x(t - 0.01) + x(t - 0.02)$$

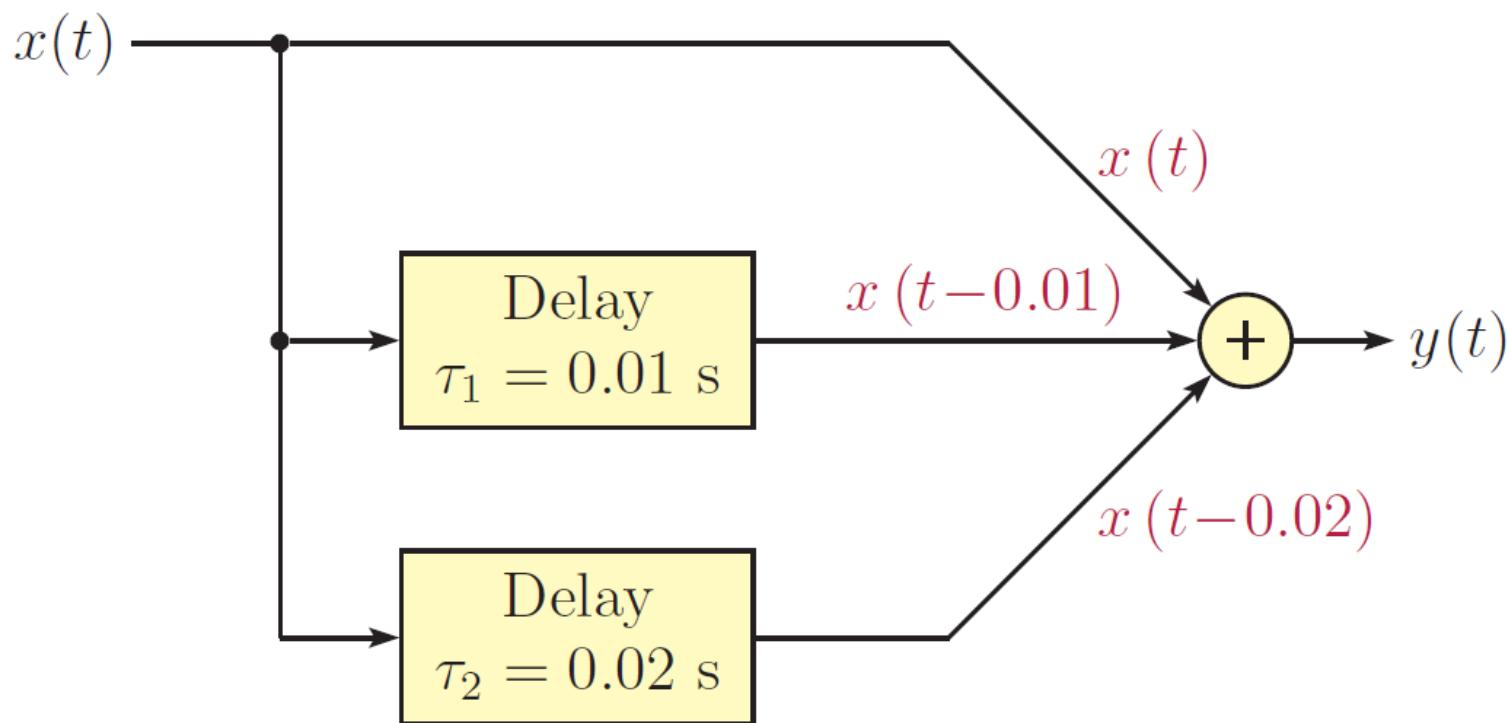
- Conversely, the system is **non-causal** since the computation of the output signal requires anticipation of a future value of the input signal.

$$y(t) = x(t) + x(t - 0.01) + x(t + 0.01)$$

Causality in Continuous-Time Systems

- The system with input-output relationship is **causal** since the **output signal** can be computed based on current and past values of the input signal.

$$y(t) = x(t) + x(t - 0.01) + x(t - 0.02)$$



Stability in Continuous-Time Systems

- A system is said to be **stable** in the bounded-input bounded-output (BIBO) sense if any **bounded input signal** is guaranteed to produce a **bounded output signal**.
- An input signal $x(t)$ is said to be **bounded** if an upper bound B_x exists for all values of t such that

$$|x(t)| < B_x < \infty \quad \text{implies that} \quad |y(t)| < B_y < \infty$$

Causality and Stability in CTLTI Systems

- The impulse response of a causal CTLTI must be equal to zero for all negative values of its argument (t).

$$h(t) = 0 \quad \text{for all } t < 0$$

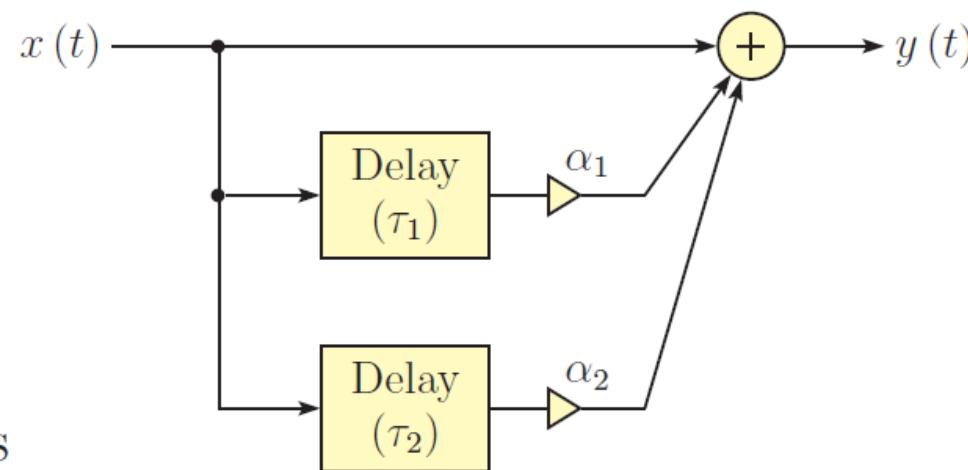
- For a CTLTI system to be stable, its impulse response must be absolute integrable.

$$\int_{-\infty}^{\infty} |h(\lambda)| d\lambda < \infty$$

Problem 2.30

2.30. The system shown in Fig. P.2.30 represents addition of echos to the signal $x(t)$:

$$y(t) = x(t) + \alpha_1 x(t - \tau_1) + \alpha_2 x(t - \tau_2)$$



Comment on the system's

- a. Linearity
- b. Time invariance
- c. Causality
- d. Stability

Problem 2.30 – Solution

- a. Let the input signal to the system be $x_1(t)$.

$$y_1(t) = \text{Sys}\{x_1(t)\} = x_1(t) + \alpha_1 x_1(t - \tau_1) + \alpha_2 x_1(t - \tau_2)$$

Similarly, if the input signal is $x_2(t)$

$$y_2(t) = \text{Sys}\{x_2(t)\} = x_2(t) + \alpha_1 x_2(t - \tau_1) + \alpha_2 x_2(t - \tau_2)$$

The response of the system to the input signal $x(t) = \beta_1 x_1(t) + \beta_2 x_2(t)$ is

$$\begin{aligned} \text{Sys}\{\beta_1 x_1(t) + \beta_2 x_2(t)\} &= \beta_1 [x_1(t) + \alpha_1 x_1(t - \tau_1) + \alpha_2 x_1(t - \tau_2)] + \beta_2 [x_2(t) + \alpha_1 x_2(t - \tau_1) + \alpha_2 x_2(t - \tau_2)] \\ &= \beta_1 y_1(t) + \beta_2 y_2(t) \end{aligned}$$

The system is linear.

Problem 2.30 – Solution

- b.** The response to $x_1(t - a)$ is

$$\text{Sys}\{x_1(t - a)\} = x_1(t - a) + \alpha_1 x_1(t - \tau_1 - a) + \alpha_2 x_1(t - \tau_2 - a) = y_1(t - a)$$

The system is time-invariant.

- c.** The system is causal provided that $\tau_1 > 0$ and $\tau_2 > 0$.
- d.** The system is stable provided that $\alpha_1, \alpha_2 < \infty$.

Problem 2.31

2.31. For each system described below determine if the system is causal and/or stable.

a. $y(t) = \text{Sys}\{x(t)\} = \int_{-\infty}^t x(\lambda) d\lambda$

b. $y(t) = \text{Sys}\{x(t)\} = \int_{t-T}^t x(\lambda) d\lambda, \quad T > 0$

c. $y(t) = \text{Sys}\{x(t)\} = \int_{t-T}^{t+T} x(\lambda) d\lambda, \quad T > 0$

Problem 2.31 (a) – Solution

a. $y(t) = \text{Sys}\{x(t)\} = \int_{-\infty}^t x(\lambda) d\lambda$

a. Let $x(t) = \delta(t)$.

$$h(t) = \text{Sys}\{\delta(t)\} = \int_{-\infty}^t \delta(\lambda) d\lambda = \begin{cases} 1, & t > 0 \\ 0, & \text{otherwise} \end{cases}$$

Therefore

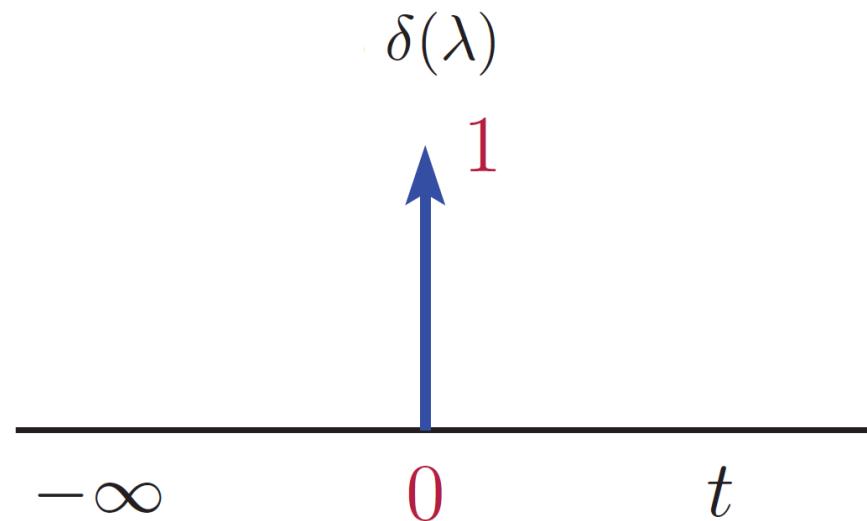
$$h(t) = u(t)$$

Since $h(t) = 0$ for $t < 0$, the system is causal. However, since $h(t)$ is not absolute summable, the system is not stable.

$$\int_0^\infty |h(\lambda)| d\lambda = \int_0^\infty d\lambda = \lambda \Big|_0^\infty = \infty$$

Problem 2.31 (a) – Explanation

a. $y(t) = \text{Sys}\{x(t)\} = \int_{-\infty}^t x(\lambda) d\lambda$



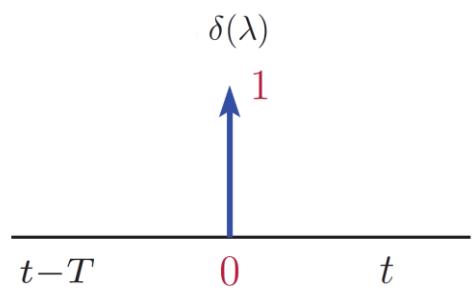
$$h(t) = \text{Sys}\{\delta(t)\} = \int_{-\infty}^t \delta(\lambda) d\lambda = \begin{cases} 1, & t > 0 \\ 0, & \text{otherwise} \end{cases}$$

Problem 2.31 (b) – Solution

b. $y(t) = \text{Sys}\{x(t)\} = \int_{t-T}^t x(\lambda) d\lambda, \quad T > 0$

b. Let $x(t) = \delta(t)$.

$$h(t) = \text{Sys}\{\delta(t)\} = \int_{t-T}^t \delta(\lambda) d\lambda = \begin{cases} 1, & 0 < t < T \\ 0, & \text{otherwise} \end{cases}$$



Therefore

$$h(t) = \Pi\left(\frac{t - T/2}{T}\right)$$

Since $h(t) = 0$ for $t < 0$, the system is causal. Also, since $h(t)$ is absolute summable, the system is stable.

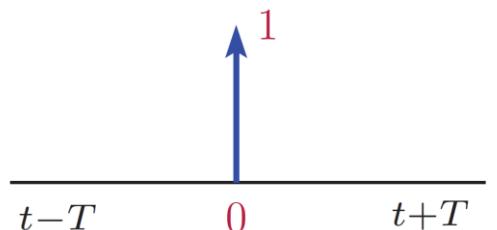
$$\int_0^\infty |h(\lambda)| d\lambda = \int_0^T d\lambda = \lambda \Big|_0^T = T$$

Problem 2.31 (c) – Solution

c. $y(t) = \text{Sys}\{x(t)\} = \int_{t-T}^{t+T} x(\lambda) d\lambda, \quad T > 0$

c. Let $x(t) = \delta(t)$.

$$h(t) = \text{Sys}\{\delta(t)\} = \int_{t-T}^{t+T} \delta(\lambda) d\lambda = \begin{cases} 1, & -T < t < T \\ 0, & \text{otherwise} \end{cases}$$



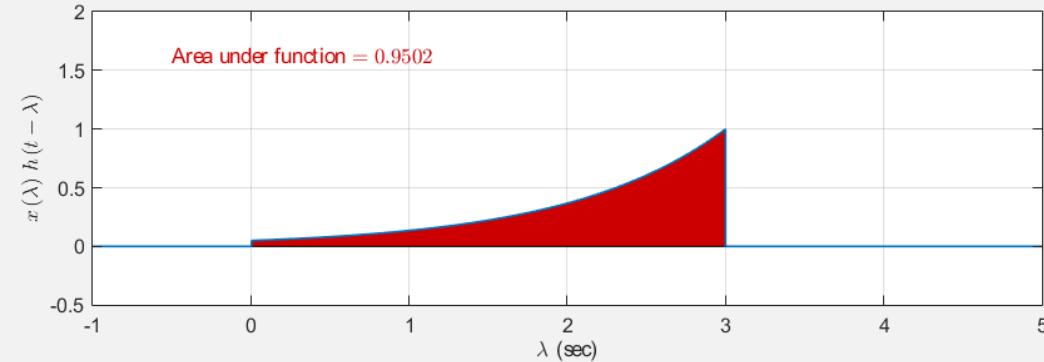
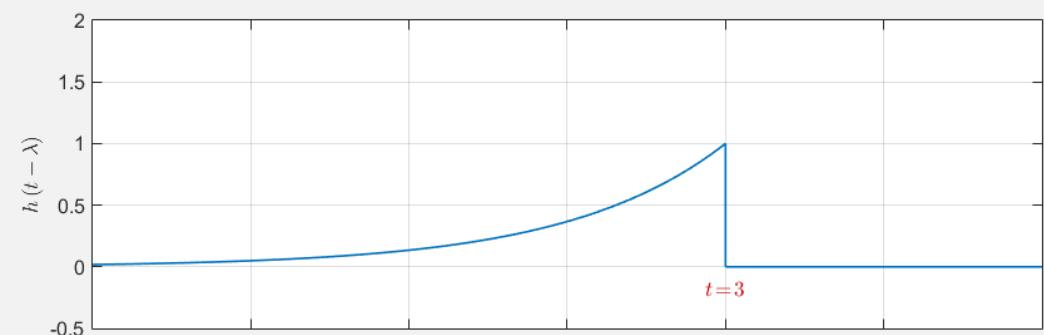
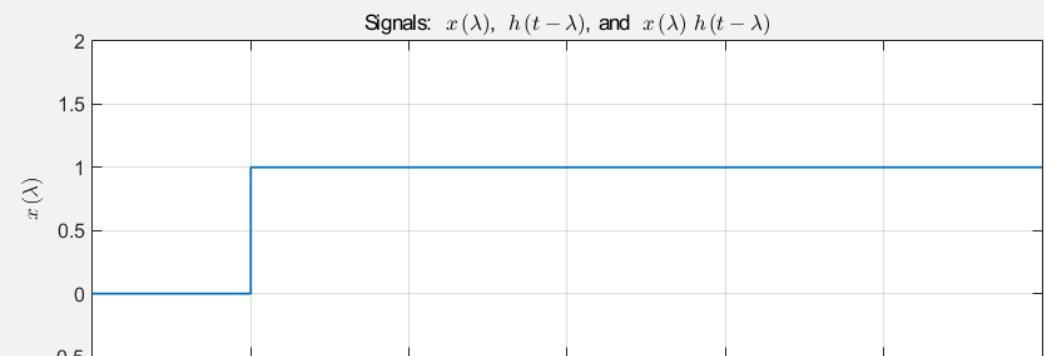
Therefore

$$h(t) = \Pi\left(\frac{t}{2T}\right)$$

Since $h(t)$ has nonzero values for some $t < 0$, the system is not causal. It is stable, however, since $h(t)$ is absolute summable.

$$\int_0^\infty |h(\lambda)| d\lambda = \int_{-T}^T d\lambda = \lambda \Big|_{-T}^T = T + T = 2T$$

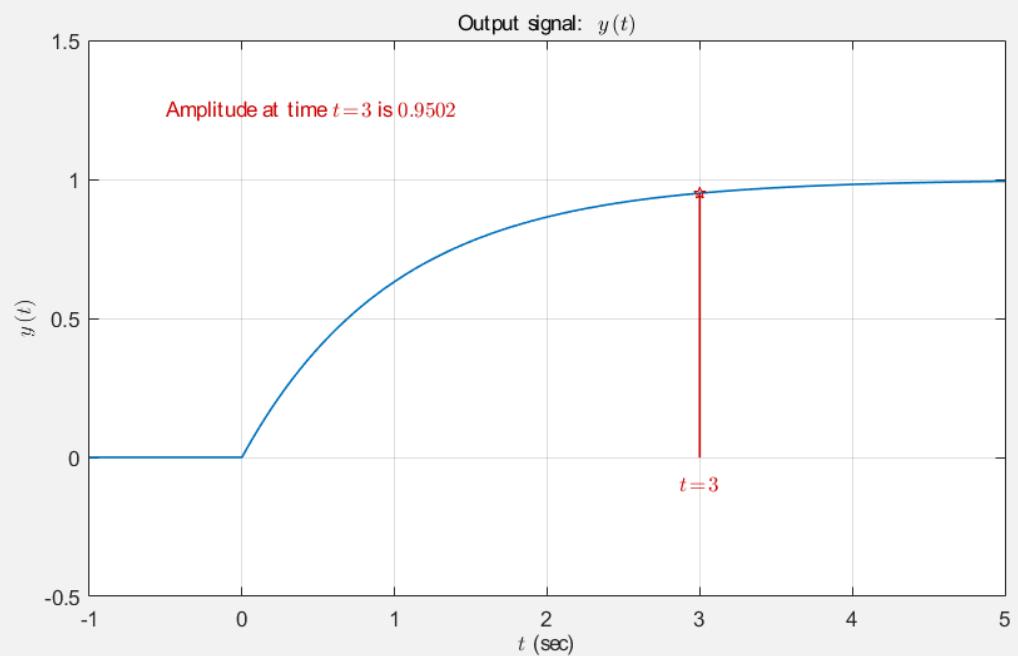
Interactive Demo: conv_demo1



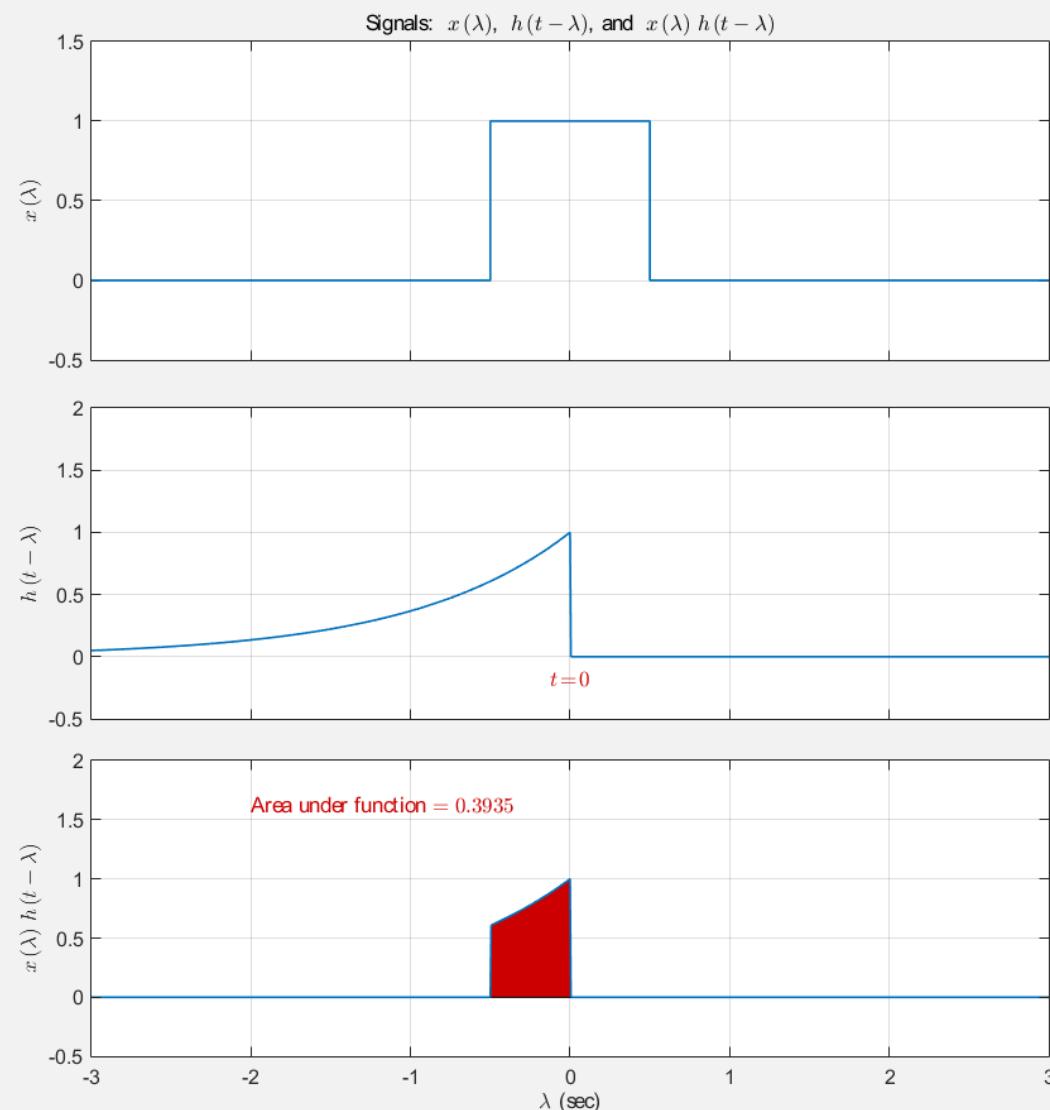
Refer to: Section 2.7.2, Pages 145 through 149,
Eqns. (2.153) and (2.154), Example 2.20,
Fig. 2.38.

Time (sec):

Time-constant (sec):



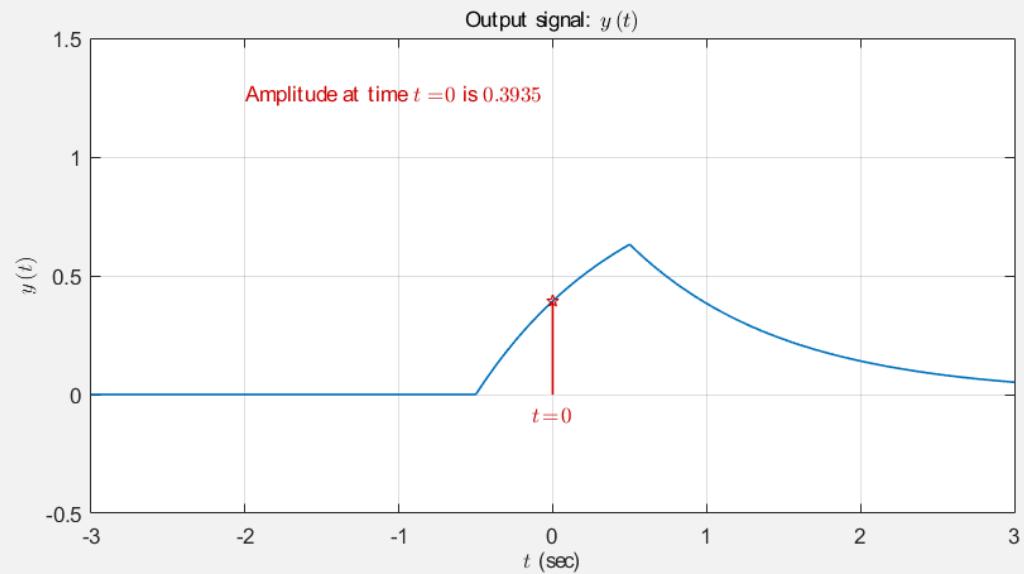
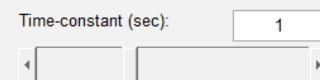
Interactive Demo: conv_demo2



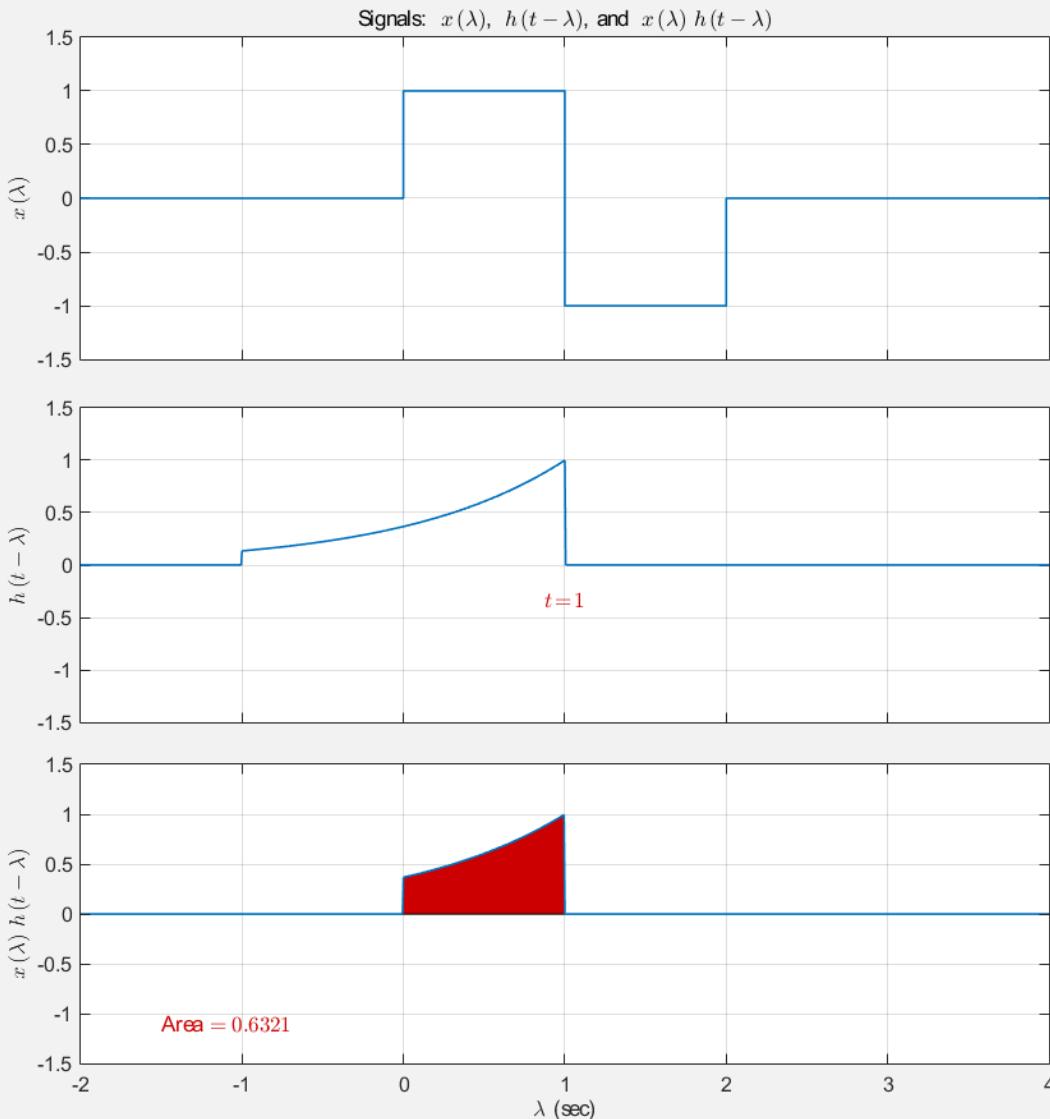
Refer to: Section 2.7.2, Pages 145 through 151,

Eqns. (2.153) and (2.154), Example 2.21,

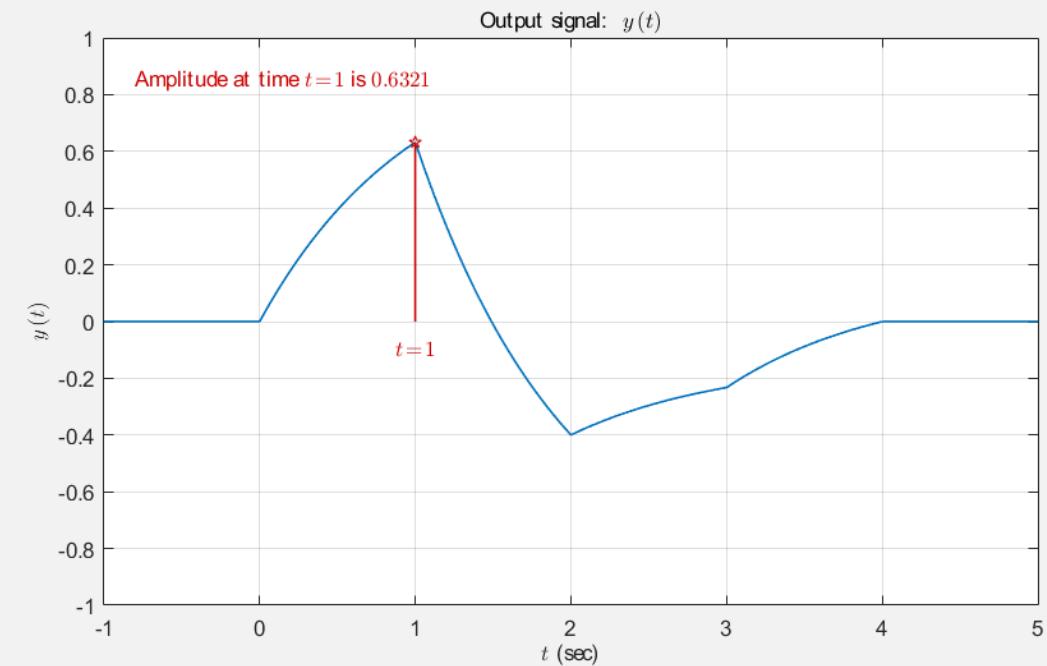
Fig. 2.39.



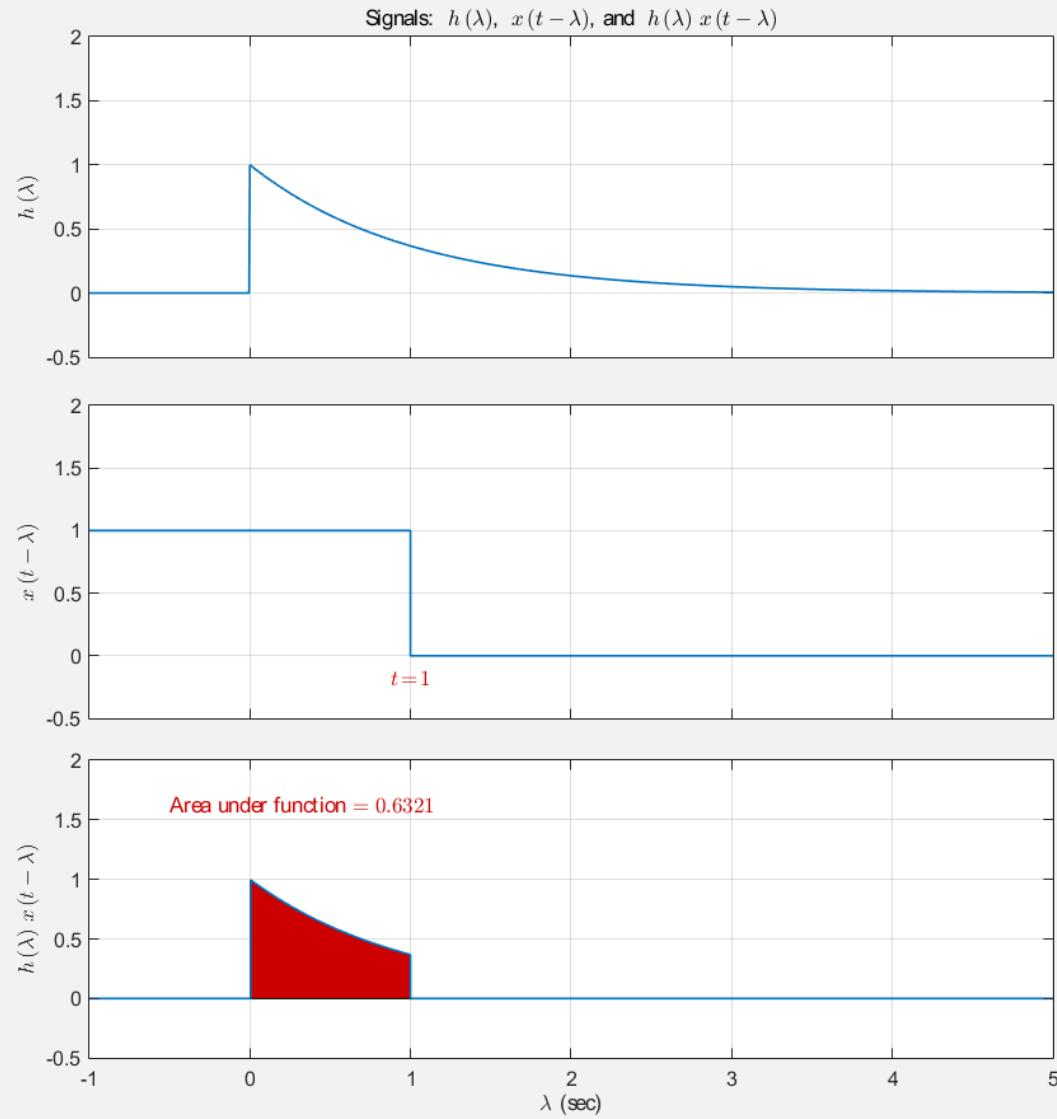
Interactive Demo: conv_demo3



Refer to: Section 2.7.2, Pages 145 through 154,
Eqns. (2.153) and (2.154), Example 2.22,
Figs. 2.41 through 2.43.



Interactive Demo: conv_demo4



Refer to: Section 2.7.2, Pages 145 through 155,

Eqns. (2.153) and (2.154), Example 2.23,

Fig. 2.44.

