

Systems

System and Systems Model

Definition 1(system): a system is an aggregation of simple physical elements according to certain topologies that achieves a defined task by transforming the input physical excitation signal to the output physical response signal.

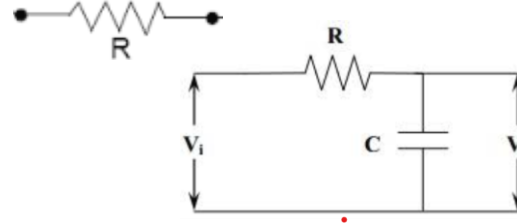
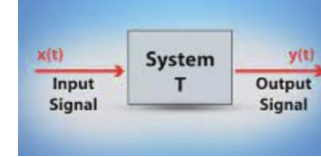
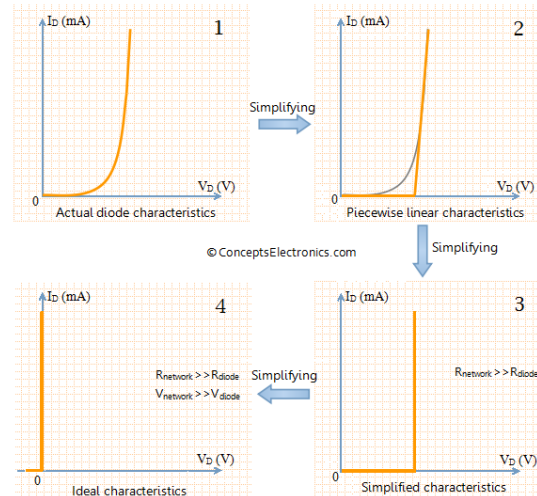
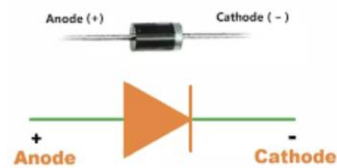
Definition 2 (system model): a system model is a mathematical function that describes the system behavior, and transformation between the input excitation signals models and the output response signals models, *under well-defined operational conditions*.

Example of system models:

An Ohmic resistor: $v(t) = Ri(t)$

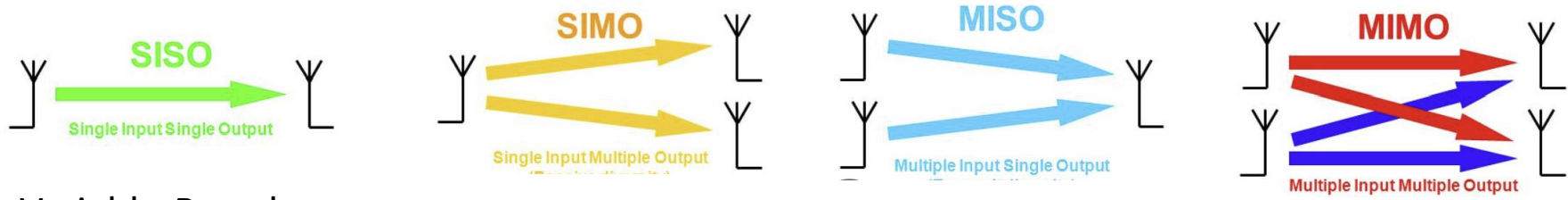
RC circuit: $\frac{dv_c(t)}{dt} + \frac{1}{RC} v_c(t) = \frac{1}{RC} v_i(t)$

A Diode:



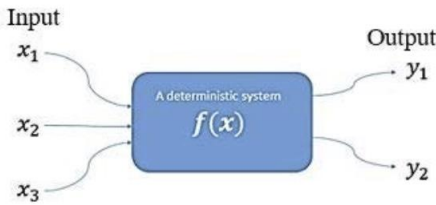
System and System Models Classification

- Input-output signals:

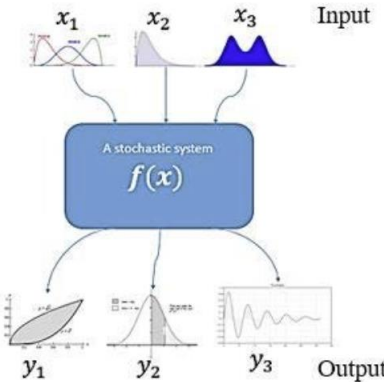


- Variable-Based:

Deterministic



Statistical



- Time-Based:

Analog: $t \in R \rightarrow y(t) \in R$

Staircase: $t \in R \rightarrow y(t) \in N$

Discrete-time $t \in N \rightarrow y(t) \in R$
Digital $t \in N \rightarrow y(t) \in N$ (with binary codes)

- Model-Based:

Linear/Nonlinear Static/Dynamic Time-Invariant/ Time-Variant Causal/Noncausal

Linear/Nonlinear System:

A system is said to be linear \leftrightarrow it satisfies the superposition principle that is,

$\forall x_1(t), x_2(t)$ inputs, α_1, α_2 parameters, and $\forall t$: if $x_1(t) \mapsto y_1(t)$, $x_2(t) \mapsto y_2(t)$ then for $x(t) = \alpha_1 x_1(t) + \alpha_2 x_2(t) \mapsto y(t) = \alpha_1 y_1(t) + \alpha_2 y_2(t)$

Linearity Test:

- additivity: if $x_1(t) \mapsto y_1(t)$, $x_2(t) \mapsto y_2(t)$
 $x(t) = x_1(t) + x_2(t) \mapsto y(t) = y_1(t) + y_2(t)$
- Proportionality: if $\bar{x}(t) \mapsto \bar{y}(t)$ then for
 $x(t) = \alpha \bar{x}(t) \mapsto y(t) = \alpha \bar{y}(t)$

Example: $v(t) = Ri(t)$

$$v_1(t) = Ri_1(t), v_2(t) = Ri_2(t)$$

Additivity: $i(t) = i_1(t) + i_2(t)$ $v(t) = R(i_1(t) + i_2(t)) = Ri_1(t) + Ri_2(t) = v_1(t) + v_2(t)$

Proportionality: $\bar{v}(t) = R\bar{i}(t) \mapsto v(t) = \alpha \bar{v}(t) = \alpha R\bar{i}(t) = \alpha \bar{v}(t)$

Example2: $v(t) = Ri(t) + i_0$

$$v_1(t) = Ri_1(t) + i_0 \quad v_2(t) = Ri_2(t) + i_0$$

Additivity: $i(t) = i_1(t) + i_2(t)$ $v(t) = R(i_1(t) + i_2(t)) + i_0 \neq v_1(t) + v_2(t)$
 $= R(i_1(t) + i_2(t)) + 2i_0 \rightarrow$ not satisfied \rightarrow Nonlinear

Exercise: determine if the system model $y(t) = e^{x(t)}$ is linear, show your proof.

Time-invariant/ Time-variant system:

Definition: A system is said to be time-invariant if it is invariant with respect to reference shift-operation. That is, if \forall excitation $x(t)$, response $y(t)$ and time shift τ , the system response to the shifted excitation $x(t - \tau)$ is a shifted response form $y(t - \tau)$.

Remark: a system model with constant parameters is a time-invariant model.

For example: the system $y(t) = 10x(t)$ is linear time-invariant and $y(t) = t\sin(t)x(t)$ linear time variant

Time invariance Test:

- Compute the system response $y(t)$ to the excitation $x(t)$ and shift the response by $\tau, y_{sh}(t)$.
- Consider the response $\bar{y}(t)$ to the new excitation $\bar{x}(t) = x(t - \tau)$
- Check if $\bar{y}(t) = y_{sh}(t)$, if yes then the system is time-invariant.

Example1:

Determine if the system $y(t) = x(2t)$ is time invariant.

Test:

- The shifted form of $y(t)$ is $y_{sh}(t) = x(2t - \tau)$
- Consider the new excitation $\bar{x}(t) = x(t - \tau)$, its response is $\bar{y}(t) = x(2(t - \tau)) = x(2t - 2\tau)$
- $\bar{y}(t) \neq y_{sh}(t)$, therefore the system is time-variant

Example2:

Determine if the system $y(t) = \sqrt{x(t)}$ is time invariant.

Test:

- The shifted form of $y(t)$ is $y_{sh}(t) = \sqrt{x(t - \tau)}$
- Consider the new excitation $\bar{x}(t) = x(t - \tau)$, its response is $\bar{y}(t) = \sqrt{x(t - \tau)}$
- $\bar{y}(t) = y_{sh}(t)$, therefore the system is time-invariant

Static and Dynamic Systems:

A system with transfer relation T is static if its response $y(t)$ occurs at the same time of its excitation $x(t)$. That is $y(t) = T[x(t)]$. (*instantaneous*)

- A static system is represented by an algebraic equation.
- A static time-invariant system is defined by a proportional relation of the type $y(t) = \alpha x(t)$

A dynamic system has a response that evolves in time based on known history or future information.

- A dynamic system is defined by a differential or integrodifferential equation.
- A dynamic time-invariant system is represented by a linear differential equation with constant coefficients.

$$\frac{d^2y}{dt^2} + 5 \frac{dy}{dt} + 5y(t) = x(t)$$

$$\frac{dy}{dt} + 5y(t) = x(t)$$

- If the differential equation is nonlinear then the system is a dynamic nonlinear system.

$$\frac{d^2y}{dt^2} + 5 \frac{dy}{dt} + 5y(t) + 6 = x(t)$$

$$\frac{dy^2}{dt} + 5y(t) = x(t)$$

- If the differential equation has variable coefficients then the system is a dynamic time variant system.

$$\frac{dy^2}{dt} + \sin(t)y(t) = x(t)$$

$$\frac{dy^2}{dt} + 5ty(t) = x(t)$$

Causal/Noncausal Systems:

Definition: A system is said to be causal if it satisfies the cause-effect principle which asserts that the response can not precede the application of the excitation.

Causal system characterization:

A system is causal if $\forall x_1(t), x_2(t)$ and $\forall \tau: x_1(t) = x_2(t) \forall t < \tau \rightarrow y_1(t) = y_2(t) \forall t < \tau$, Or alternatively, if $\forall x(t)$, $x(t) = 0 \forall t < \tau \rightarrow y(t) = 0 \forall t < \tau$

Test:

For causality Check: prove that $t_{excitation} < t_{response}$

For noncausality check: find a case at which $t_{excitation} > t_{response}$

Example1:

Determine if the systems $y(t) = x(t - \tau)$ is causal for:

- $\tau > 0$
- $\tau < 0$
- Solution: for the system to be causal the system model must satisfy $t - \tau \leq t$, that is $-\tau \leq 0 \rightarrow \tau \geq 0$

Example2:

Determine if the system $y(t) = x(\sqrt{t})$ is causal.

Solution: $\sqrt{t} \leq t \rightarrow 0 \leq t \leq t^2 \rightarrow 0 < 1 \leq t$, so the system is not causal-->

In fact $y\left(\frac{1}{4}\right) = x\left(\frac{1}{2}\right)$

Linear Time Invariant Response (LTI):

Impulse Response:

Definition: The zero-state response $h(t)$ of a linear time-invariant system with transform $y(t) = T[x(t)]$ for the excitation input $\delta(t)$ is said to be the impulse response of the system.

The impulse response completely characterizes the LTI system. Moreover, the zero state response $y(t)$ of an LTI system, with impulse response $h(t)$, to any excitation signal $x(t)$ system can be computed using the convolution integral of $x(t)$, and $h(t)$ (discussed later)

Determination of the impulse response and the solution of a dynamic LTI system to any singularity signal excitation:

Procedure:

- Determine the zero-input response $g(t)$ of the n^{th} order dynamic system for $t \geq 0^+$
- Build the response model using the form: $y(t) = g(t)u(t) + \sum_{k=0}^m \alpha_k u_k(t)$ with u_m the minimum order singularity signal with n^{th} derivative that covers the maximum order excitation singularity signal term.
- Apply the generalized identity of singularity signals to construct the relations that defines α_k in terms of $g(t)$ and its derivatives at $t = 0$.
- Equate the expressions of $g(t)$ and its derivatives to the values obtained from the identity of singularity signals.
- Solve the set of equations to determine the values of the parameters of the solution $g(t)$ and the α_k s

Example1:

Determine the impulse response of the system $\frac{dy(t)}{dt} + 5y(t) = 2x(t)$.

Solution: The impulse response $h(t)$ is obtained to $x(t) = \delta(t)$, thus the equation becomes:

$$\frac{dh(t)}{dt} + 5h(t) = 2\delta(t)$$

The zero input response of the system is the solution of $\frac{dg(t)}{dt} + 5g(t) = 0$ which is given by:

$$g(t) = Ae^{-5t}$$

The solution $h(t)$ is constructed as $h(t) = g(t)u(t)$, since its first derivative (system order) has the term $\delta(t)$ that is all the parameters $\alpha_k = 0$.

Now compute the derivative of $h(t)$ and apply in the differential equation (**remember that $g(t)\delta(t) = g(0)\delta(t)$**):

$$g'(t)u(t) + g(0)\delta(t) + 5g(t)u(t) = 2\delta(t) \leftrightarrow g(0) = 2.$$

$$\text{Applying } g(0^+) = 2 = Ae^{+0} \rightarrow A = 2.$$

The solution is: $h(t) = 2e^{-5t}u(t)$

Exercise: compute the solution of the same system for $x(t) = 10\delta(2t - 8)$.

Example2: Determine the response of the system $\frac{dy(t)}{dt} + 5y(t) = 2\dot{x}(t)$ for $x(t) = \delta(t)$.

Solution: The zero input response of the system is the solution of $\frac{dg(t)}{dt} + 5g(t) = 0$ which is given by: $g(t) = Ae^{-5t}$

The solution $y(t)$ is constructed as $y(t) = g(t)u(t) + B\delta(t)$, since its first derivative (system order) has the term $\dot{\delta}(t)$. In fact with out adding the zero order singularity signal $\delta(t)$ the maximum order derivative (first order in this case) has the maximum singularity order $\delta(t)$ which does not cover the $\dot{\delta}(t)$ term of the right side.

Differentiating $y(t)$ and applying in the differential equation we get

$$g'(t)u(t) + g(0)\delta(t) + B\dot{\delta}(t) + 5g(t)u(t) + 5B\delta(t) = 2\dot{\delta}(t) \leftrightarrow B = 2 \text{ and } g(0) + 5B = 0 \rightarrow g(0) = -10 = Ae^0 = A.$$

Thus the solution is given by: $y(t) = -10e^{-5t}u(t) + 2\delta(t)$

Exercise: Determine the response of the system $\frac{dy(t)}{dt} + 5y(t) = 2\ddot{x}(t)$ to $x(t) = \delta(t)$.

Important fact and exercise: Observe that the zero-state response to the derivative of $\delta(t)$ is the derivative of the response to $\delta(t)$. This result is generalized in the following theorem(show).

Theorem (zero state response):

Given an LTI system with zero state response $y(t)$ to the input excitation $x(t)$, then the zero-state response to:

- $\frac{dx(t)}{dt}$ is $\frac{dy(t)}{dt}$
- $\int_0^t x(\sigma)d\sigma$ is $\int_0^t y(\sigma)d\sigma$

Example3:

Determine the response of the system $\frac{d^2y}{dt^2} + 3\frac{dy}{dt} + 2y(t) = \ddot{x}(t)$, for $x(t) = \delta(t)$.

- The system has characteristic roots of the characteristic algebraic equation: $\sigma_1 = -1$ and $\sigma_2 = -2$, therefore the zero input response $g(t) = Ae^{-t} + Be^{-2t}$
- The zero-state solution is $y(t) = g(t)u(t) + C\delta(t)$
- $2 \times y(t) = 2 \times (g(t)u(t) + C\delta(t))$
- $3 \times \frac{dy(t)}{dt} = 3 \times (g'(t)u(t) + g(0)\delta(t) + C\delta'(t))$
- $\frac{d^2y}{dt^2} = g''(t)u(t) + g'(0)\delta(t) + g(0)\delta'(t) + C\delta''(t)$
- Identity of singularity signals:
 - Balance of δ'' : $C=1$
 - Balance of δ' : $g(0) + 3C = 0 \leftrightarrow g(0) = -3 \times 1 = -3$
 - Balance of δ : $g'(0) + 3g(0) + 2C = 0 \leftrightarrow g'(0) = -3 \times -3 - 2 \times 1 = 7$

Computing A and B using $g(0)$ and $g'(0)$:

$$g(0^+) = -3 = A + B$$

$$g'(0) = 7 = -A - 2B$$

Solving the system we obtain $B = -4$ and $A = -3 + 4 = 1$. Thus the solution is:

$$y(t) = (e^{-t} - 4e^{-2t})u(t) + \delta(t)$$

Exercise:

Compute the response to $x(t) = \dot{\delta}(t)$ and then compute it from the solution of Example 3 using the zero-state response theorem, compare.

Convolution Integral:

Theorem: Impulse and step response of an LTI system:

Given an LTI system with impulse response $h(t)$ and/or a step response $a(t)$, the zero state response of the system to any input $x(t)$ can be determined by the convolution integrals:

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

$$y(t) = x'(t) * a(t) = \int_{-\infty}^{\infty} x'(\sigma) a(t - \sigma) d\sigma \quad (\text{Duhamel's Integral})$$

Proof (convolution with $h(t)$): from the convolution property of $\delta(t)$ we can write:

$x(t) = \int_{-\infty}^{\infty} x(\sigma) \delta(t - \sigma) d\sigma$, applying the linear time invariant transform T to this excitation input we obtain

$$y(t) = T[x(t)] = T\left[\int_{-\infty}^{\infty} x(\sigma) \delta(t - \sigma) d\sigma\right] = \int_{-\infty}^{\infty} x(\sigma) T[\delta(t - \sigma)] d\sigma$$

$$= \int_{-\infty}^{\infty} x(\sigma) h(t - \sigma) d\sigma$$

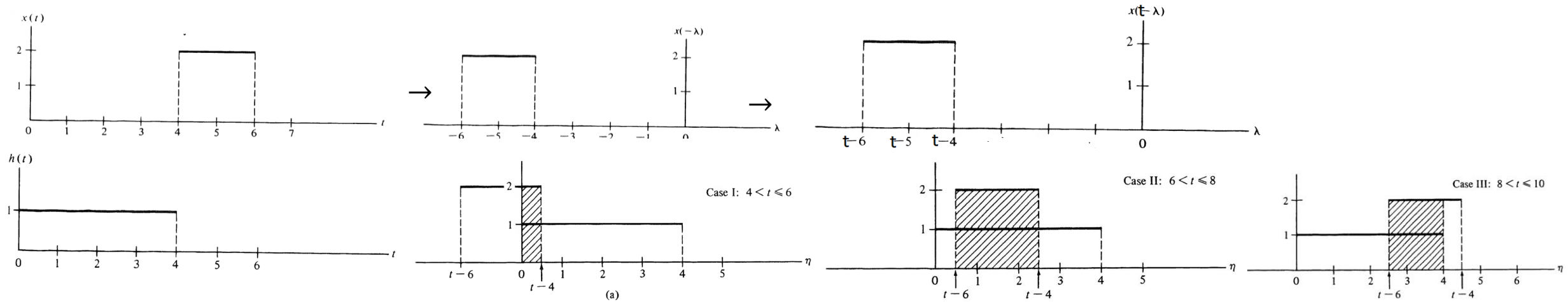
Computation of convolution Integral $x_1(t) * x_2(t)$:

- Transfer of the independent variable from the t space to the σ space $x_1(t) \rightarrow x_1(\sigma), x_2(t) \rightarrow x_2(\sigma)$.
- folding of one of the two inputs of the convolution operator $x_2(\sigma) \rightarrow x_2(-\sigma)$.
- Shift of the folded variable by t : $x_2(-\sigma) \rightarrow x_2(t - \sigma)$
- Integration of the multiplication using the form $\int_{-\infty}^{\infty} x_1(\sigma) x_2(t - \sigma) d\sigma$ over the integrand definition ranges

Properties of the convolution operator:

- $x_1(t) * x_2(t) = x_2(t) * x_1(t)$
- $x_1(t) * [\alpha x_2(t)] = \alpha [x_1(t) * x_2(t)]$
- $x_1(t) * [x_2(t) + x_3(t)] = x_1(t) * x_2(t) + x_1(t) * x_3(t)$
- $x_1(t) * [x_2(t) * x_3(t)] = [x_1(t) * x_2(t)] * x_3(t)$
- If $x_1(t)$ is time limited to $[a \ b]$, $x_2(t)$ is time limited to $[c \ d]$, then $x_1(t) * x_2(t)$ is time limited to $[a + c \ b + d]$
- If the area under $x_1(t)$ is A_1 and the area under $x_2(t)$ is A_2 , then the area under $x_1(t) * x_2(t)$ is $A_1 \times A_2$

Example: Compute the response of the LTI system with impulse response $h(t) = \pi(\frac{t-2}{4})$ to the input $x(t) = 2\pi(\frac{t-5}{2})$

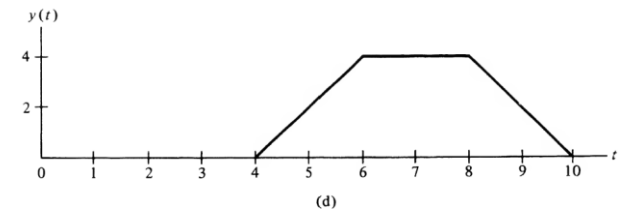


$$y(t) = 0 \text{ for } t \leq 4 \text{ and } t \geq 10$$

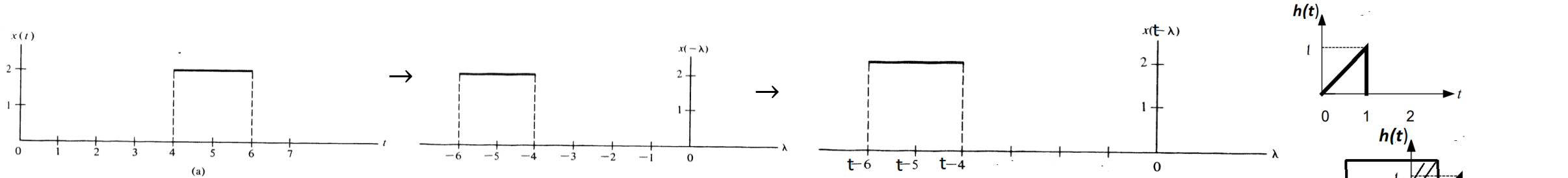
$$y(t) = \int_0^{t-4} 2 \cdot 1 d\sigma = 2(t-4) \text{ for } t-4 \leq 4 \text{ and } t-4 \geq 0 \text{ and } t-6 \leq 0 \rightarrow t \in [4, 6]$$

$$y(t) = \int_{t-6}^{t-4} 2 \cdot 1 d\sigma = 4 \text{ for } t-4 \leq 4 \text{ and } t-4 \geq 0 \text{ and } t-6 \geq 0 \rightarrow t \in [6, 8]$$

$$y(t) = \int_{t-6}^4 2 \cdot 1 d\sigma = 2(10-t) \text{ for } t-4 \geq 4 \text{ and } t-6 \geq 0 \text{ and } t-6 \leq 4 \rightarrow t \in [8, 10]$$



Example: Compute the response of the LTI system with impulse response $h(t) = r(t) \pi(\frac{t-1}{2})$ to the input $x(t) = 2\pi(\frac{t-5}{2})$

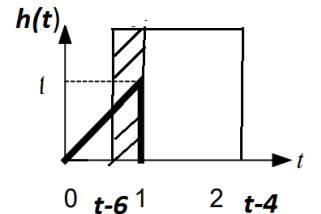
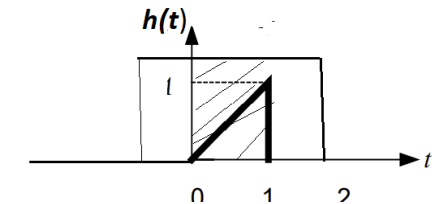
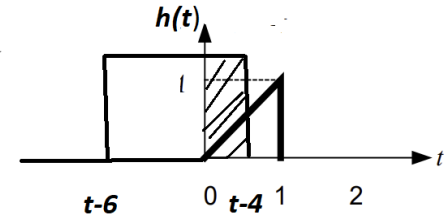


$y(t) = 0$ for $t \leq 4$ and $t \geq 7$

$$y(t) = \int_0^{t-4} 2 \times \sigma d\sigma = (t-4)^2 \quad \text{for} \quad \begin{cases} t-4 \geq 0 \\ t-4 \leq 1 \\ t-6 \leq 0 \end{cases} \rightarrow 4 \leq t \leq 5$$

$$y(t) = \int_0^1 2 \times \sigma d\sigma = 1 \quad \text{for} \quad \begin{cases} t-4 \geq 0 \\ t-4 \geq 1 \\ t-6 \leq 0 \end{cases} \rightarrow 5 \leq t \leq 6$$

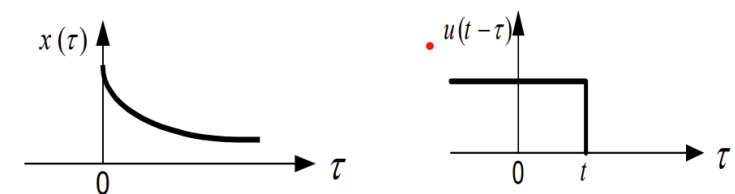
$$y(t) = \int_{t-6}^1 2 \times \sigma d\sigma = 1 - (t-6)^2 \quad \text{for} \quad \begin{cases} t-4 \geq 1 \\ t-6 \geq 0 \\ t-6 \leq 1 \end{cases} \rightarrow 6 \leq t \leq 7$$



Example: Compute the response of the LTI system with impulse response $h(t) = u(t)$ to the input $x(t) = e^{-2t}u(t)$

$y(t) = 0$ for $t \leq 0$

$$y(t) = \int_0^t e^{-2\tau} d\tau = \frac{e^{-2t} - 1}{-2} = \frac{1 - e^{-2t}}{2} \quad \text{for} \quad t \geq 0$$



Sinusoidal Steady State Response:

Theorem: Given an LTI system with impulse response $h(t)$, the response of the system to a sinusoidal input $x(t) = X\cos(\omega_0 t + \varphi)$ is sinusoidal with the same input frequency $y(t) = Y\cos(\omega_0 t + \theta)$ with:

$$Y = X \cdot |H(\omega)|_{\omega=\omega_0}, \quad \theta = \varphi + \angle H(\omega)|_{\omega=\omega_0}$$

Proof:

let $x(t) = X e^{j(\omega_0 t + \varphi)}$ then $y(t) = \int_{-\infty}^{\infty} h(\tau) \cdot X e^{j(\omega_0(t-\tau) + \varphi)} d\tau = X e^{j(\omega_0 t + \varphi)} \int_{-\infty}^{\infty} h(\tau) \cdot e^{-j(\omega_0 \tau)} d\tau = X e^{j(\omega_0 t + \varphi)} \cdot H(\omega)|_{\omega_0}$

Where $H(\omega) = \int_{-\infty}^{\infty} h(\tau) \cdot e^{-j\omega\tau} d\tau$ is the frequency response of the system the characterizes the spectral response of the linear time invariant system. $H(\omega)$ is a complex function of the real variable ω that represents the Fourier transform of the impulse response $h(t)$.

$Re(x(t)) = X\cos(\omega_0 t + \varphi) \rightarrow Re(y(t)) = Y\cos(\omega_0 t + \theta)$ which proves the assertion of the theorem

Example1: compute the frequency response of the system with impulse response $h(t) = 10e^{-2t}u(t)$

Solution: $H(\omega) = \int_{-\infty}^{\infty} h(\tau) \cdot e^{-j\omega\tau} d\tau = \int_0^{\infty} 10e^{-2\tau} \cdot e^{-j\omega\tau} d\tau = \int_0^{\infty} 10e^{-(2+j\omega)\tau} d\tau = 10 \frac{e^{-(2+j\omega)\tau}}{-(2+j\omega)} \Big|_0^{\infty} = 10 \frac{1}{(2+j\omega)}$

$$|H(\omega)| = \frac{10}{\sqrt{4 + \omega^2}}, \quad \angle H(\omega) = -\tan^{-1}\left(\frac{\omega}{2}\right)$$

Example2(sinusoidal steady-state response): compute the steady-state response of the system in example1 to the input signal

$$x(t) = 2 \cos(4t + \frac{\pi}{3}) + 5 \sin(6t + \frac{\pi}{4})$$

Solution: the input is composed of two sinusoidal signals so we can apply superposition and compute the sinusoidal steady-state response of each sinusoid using the theorem.

$$y(t) = 2 \cdot \frac{10}{\sqrt{4 + 4^2}} \cos(4t + \frac{\pi}{3} - \tan^{-1}\left(\frac{4}{2}\right)) + 5 \cdot \frac{10}{\sqrt{4 + 6^2}} \sin(6t + \frac{\pi}{4} - \tan^{-1}\left(\frac{6}{2}\right)) =$$
$$= \frac{20}{\sqrt{20}} \cos(4t + \frac{\pi}{3} - \tan^{-1}(2)) + \frac{50}{\sqrt{40}} \sin(6t + \frac{\pi}{4} - \tan^{-1}(3))$$

(compute the final form, note that the argument of the \tan^{-1} is in radiant)

System Stability:

Definition: An LTI system is said to be asymptotically stable if its transient response goes to zero and a steady state response is reached for t goes to infinity.

Theorem1: an LTI system with impulse response $h(t)$ is asymptotically stable $\leftrightarrow \lim_{t \rightarrow \infty} h(t) = 0$.

Theorem2: adynamic LTI system is asymptotically stable \leftrightarrow all the roots of its characteristic equation/ the poles of its transfer function have a negative real part (located in the left semi plan of the complex plan)

Theorem3: an LTI system is unstable if it has at least a positive real-part root or a repeated root with zero real part.

Definition:(BIBO stability) an LTI system is said to be BIBO (Bounded Input/Bounded Output) $\leftrightarrow \forall \text{ input } x(t) \text{ with } |x(t)| \leq N, \exists M < \infty$ so that the respons $|y(t)| \leq M, \forall t$ (weak stability)

Theorom4: a system is BIBO stable $\leftrightarrow \int_{-\infty}^{\infty} |h(t)| dt < \infty$ that is if its impulse response is absolutely integrable.

Exercise: prove this theorem.

Theorom5: a system is BIBO stable if it has no roots with positive real parts and all the roots with zero real part are not repeated roots.

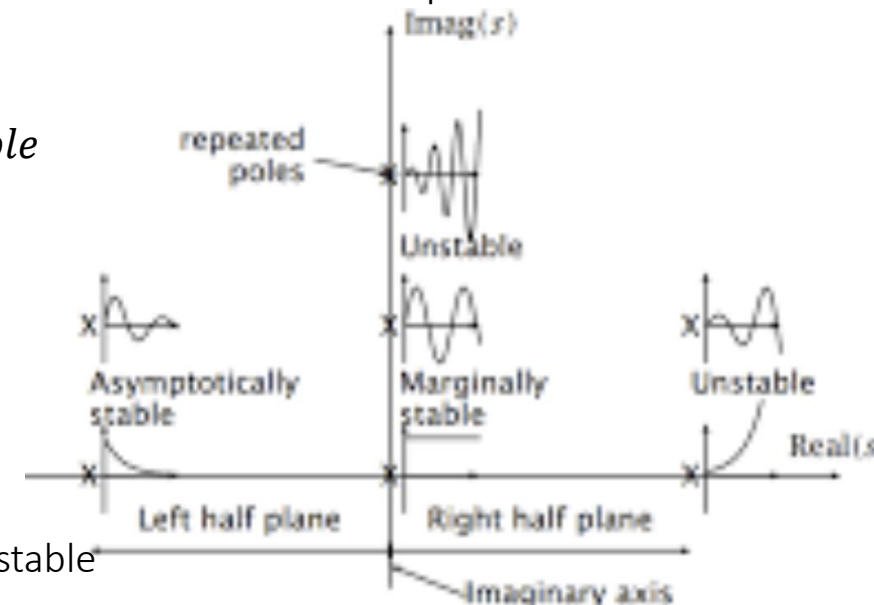
Example1: discuss the stability of the following dynamic systems:

- $\frac{d^2y}{dt^2} + 3\frac{dy}{dt} + 2y(t) = x(t)$, the roots of are $\sigma_1 = -1, \sigma_2 = -2 \rightarrow$ *asymptotically stable*
- $H(s) = \frac{(s+4)}{(s^2+4)(s+1)} \rightarrow$ *BIBO stable*
- $H(s) = \frac{(s+4)}{(s^2+4)^2(s+1)} \rightarrow$ *Unstable*
- $H(s) = \frac{(s+4)}{(s+2)(s-1)} \rightarrow$ *Unstable*

Asymptotically stable \rightarrow BIBO Stable

Example2: Prove that the system with the following $h(t)$ achieves the BIBO stability theorem

$$h(t) = 10e^{-3t}u(t) \rightarrow \int_{-\infty}^{\infty} |10e^{-3t}u(t)| dt \rightarrow \int_0^{\infty} 10e^{-3t} dt = \frac{10e^{-3t}}{-3} \Big|_0^{\infty} = \frac{10}{3} < \infty \rightarrow \text{BIBO stable}$$



Modeling and Simulation of an LTI System:

Modeling a system for simulation and prototyping purposes means constructing an internal representation (state space representation) for a given external model representation (differential equation/ Laplace transform). While the external model is unique, the internal model is not. The selected internal topology should serve the simulation or prototyping objectives.

Simulation: Using computer packages (such as Matlab, Mathcad, LabView,...) to analyze system characteristics and its response to various excitation input signals.

Prototyping: Building a system model using hardware components for testing and analysis objectives.

Observer Representation Model:

As an example of modeling, we consider the observer representation which can be defined by separate and integrate processes.

Example1: Determine the observer model of the system defined by:

$$\frac{d^4y}{dt^4} - 5\frac{d^3y}{dt^3} + 2\frac{d^2y}{dt^2} + 3\frac{dy}{dt} + y = 6\frac{d^3x}{dt^3} + 4\frac{d^2x}{dt^2} + 7\frac{dx}{dt} + 2x$$

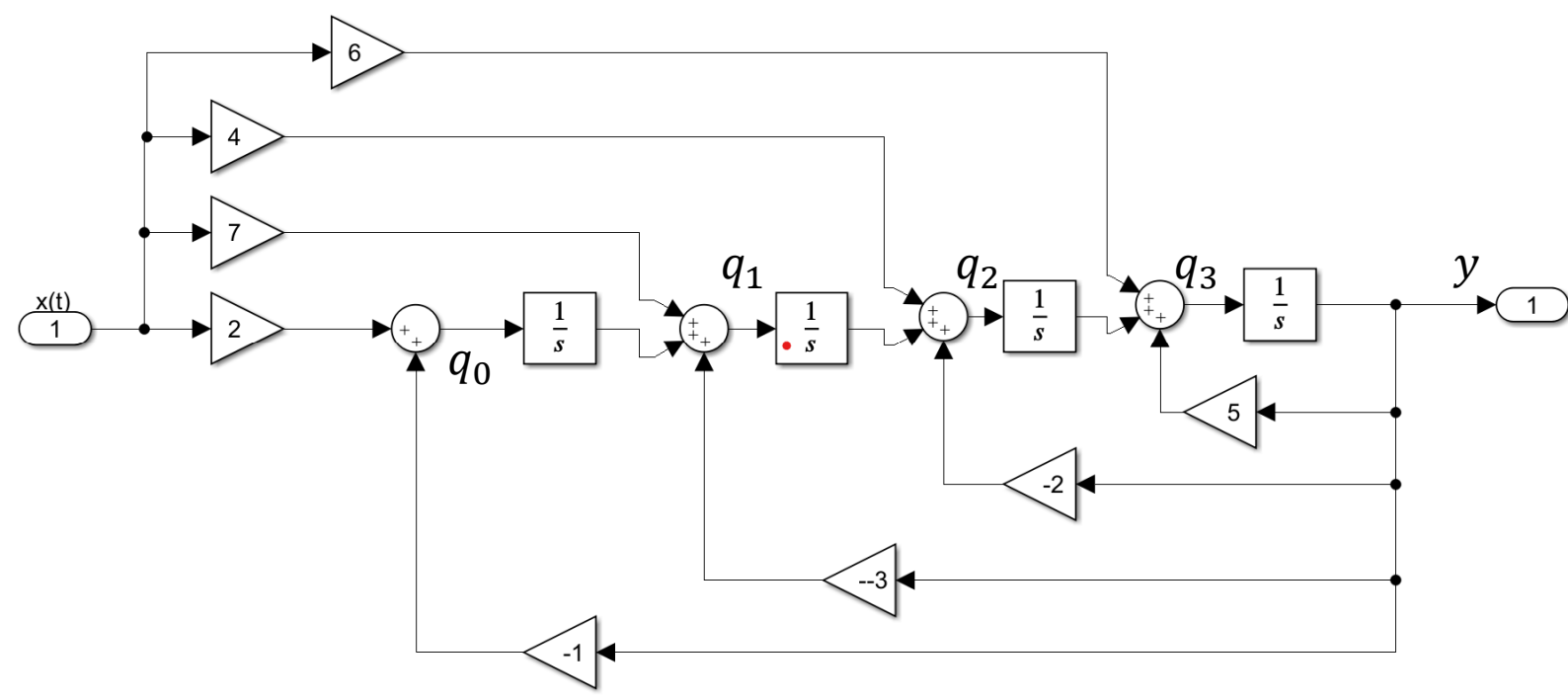
Separate: $\frac{d^4y}{dt^4} - 5\frac{d^3y}{dt^3} + 2\frac{d^2y}{dt^2} + 3\frac{dy}{dt} - 6\frac{d^3x}{dt^3} - 4\frac{d^2x}{dt^2} - 7\frac{dx}{dt} = 2x - y = q_0$

Integrate + Separate: $\frac{d^3y}{dt^3} - 5\frac{d^2y}{dt^2} + 2\frac{dy}{dt} - 6\frac{d^2x}{dt^2} - 4\frac{dx}{dt} = \int_0^t q_0 d\sigma + 7x - 3y = q_1$

Integrate + Separate: $\frac{d^2y}{dt^2} - 5\frac{dy}{dt} - 6\frac{dx}{dt} = \int_0^t q_1 d\sigma + 4x - 2y = q_2$

Integrate + Separate: $\frac{dy}{dt} = \int_0^t q_2 d\sigma + 6x + 5y = q_3$

Integrate: $y = \int_0^t q_3 d\sigma$



Example2: Determine the observer model of the system defined by:

$$\frac{d^4 y}{dt^4} + 2 \frac{d^2 y}{dt^2} + 3 \frac{dy}{dt} + 5y = 3 \frac{d^3 x}{dt^3} + 4 \frac{dx}{dt} + 2x$$

Separate: $\frac{d^4 y}{dt^4} - 5 \frac{d^3 y}{dt^3} + 2 \frac{d^2 y}{dt^2} + 3 \frac{dy}{dt} - 6 \frac{d^3 x}{dt^3} - 4 \frac{d^2 x}{dt^2} - 7 \frac{dx}{dt} = 2x - 5y = q_0$

Integrate + Separate: $\frac{d^3 y}{dt^3} + 3 \frac{dy}{dt} - 6 \frac{d^2 x}{dt^2} = \int_0^t q_0 d\sigma + 4x - 3y = q_1$

Integrate + Separate: $\frac{d^2 y}{dt^2} - 6 \frac{dx}{dt} = \int_0^t q_1 d\sigma - 2y = q_2$

Integrate + Separate: $\frac{dy}{dt} = \int_0^t q_2 d\sigma + 3x = q_3$

Integrate: $y = \int_0^t q_3 d\sigma$

