# Signals and Systems



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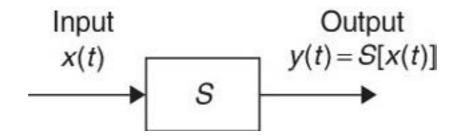
Linear and time-invariant systems
The impulse response and the convolution integral
Linear ordinary differential equations and LTI systems
Causality
BIBO stability

#### Book:

Chapter 2

**Exercises:** 2.1, 2.2, 2.3, 2.4, 2.5, 2.8, 2.9, 2.11 (3rd Ed.) 2.1, 2.2, 2.5, 2.7, 2.8, 2.12, 2.14, 2.18 (2nd Ed.)

- We consider a system with a single input and a single output
- Input signal: x(t), output signal y(t)
- Such systems are called SISO systems
- SISO stands for Single Input Single Output
- If a system has Multiple Inputs and Multiple Outputs it is called a MIMO system
- We restrict ourselves to SISO systems



- The action of the system on the input signal x(t) is described by the system operator S
- We write

$$y(t) = S\{x(t)\}$$

- In this course we are particularly interested in systems that are Linear and Time-Invariant
- Such systems are called LTI systems

• **Linearity** Suppose we have two input signals  $x_1(t)$  and  $x_2(t)$ . Denote the corresponding output signals by  $y_1(t)$  and  $y_2(t)$ :

$$y_1(t) = S\{x_1(t)\}$$
 and  $y_2(t) = S\{x_2(t)\}$ 

The system is called *linear* if

$$y(t) = S\{\alpha x_1(t) + \beta x_2(t)\}\$$
  
=  $\alpha S\{x_1(t)\} + \beta S\{x_2(t)\}\$   
=  $\alpha y_1(t) + \beta y_2(t)$ 

for any two constants  $\alpha$  and  $\beta$ 

- Any linear combination of input signals produces the same linear combination of their corresponding output signals
- Taking  $\beta = 0$ , it follows from the above definition that

$$y(t) = S\{\alpha x_1(t)\} = \alpha S\{x_1(t)\} = \alpha y_1(t)$$

• In other words, if you scale the input signal by a factor  $\alpha$ , the output signal will scale with the same factor

• **Example** Consider a SISO system with input signal x(t) and an output signal given by

$$y(t) = \frac{1}{T} \int_{\tau=t-T}^{t} x(\tau) d\tau + B,$$

where *B* is a constant. Such a system is called a biased averager (can you see why?)

• Scaling the input signal by a factor  $\alpha$ , we obtain the output signal

$$\frac{\alpha}{T} \int_{\tau=t-T}^{t} x(\tau) \, \mathrm{d}\tau + B,$$

which is not equal to  $\alpha y(t)$  unless B = 0

- The averager is nonlinear for  $B \neq 0$
- For B = 0, it is easy to see that a linear combination of input signals produces the same linear combination of the corresponding output signals
- The averager is linear for B = 0

• **Time-Invariance** Let y(t) be the output signal that corresponds to an input signal x(t):

$$y(t) = S\{x(t)\}$$

• The system is called *time-invariant* if

$$y(t \pm \tau) = S\{x(t \pm \tau)\}\$$

for any time shift  $\tau > 0$ 

• In words: shifting your input signal produces an equally time-shifted output signal

- Let the Dirac distribution be the input signal of an LTI system
- The corresponding output signal is written as h(t) and is called the im- $pulse\ response$ :

$$h(t) = S\{\delta(t)\}$$

• We claim that if you know the impulse response of an LTI system then you know the response to any other input signal!

• To show this, let y(t) be the output signal that corresponds to an input signal x(t):

$$y(t) = S\{x(t)\}$$

• Because of the sifting property of the Dirac distribution, we have

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

• The right-hand side of the above expression can be seen as a continuous weighted summation of shifted Dirac distributions

• Substitution gives

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

- Now note that *S* is linear and acts on functions that depend on time *t*
- This allows us to write

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

• Since the system is time-invariant as well, we have

$$h(t-\tau) = S\{\delta(t-\tau)\}\$$

and we arrive at

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

• Knowing the impulse response h(t), we can determine the response y(t) to any input signal x(t) by evaluating the above integral

- This integral is called a *convolution integral*
- Short-hand notation:

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

- The asterisk is called the *convolution product*
- The output signal y(t) is equal to the convolution product of the input signal x(t) and the impulse response h(t)

• For two real numbers *a* and *b*, we have

$$ab = ba$$

- The product of two real numbers commutes
- Is this also true for the convolution product? In other words, do we have

$$x * h = h * x$$
?

• The answer is yes. Let's check it.

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

$$= \int_{\tau = -\infty}^{\infty} x(\tau)h(t - \tau) d\tau \stackrel{p = t - \tau}{=} \int_{p = -\infty}^{\infty} x(t - p)h(p) dp$$

$$= \int_{p = -\infty}^{\infty} h(p)x(t - p) dp = h * x$$

• Conclusion: the convolution product of two signals commutes (due to the minus sign in the argument of *h*)

• If you change the minus sign into a plus sign you get what is called the *cross-correlation* of the two signals x(t) and h(t) provided these signals are both real-valued:

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

• The cross correlation of two signals does *not* commute

$$x \star h \neq h \star x$$

• For the product of real numbers, there exists an identity element called "one" and written as 1 for which

$$a = a \cdot 1 = 1 \cdot a$$

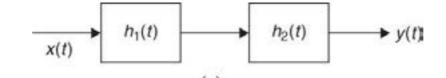
- What is the identity element for the convolution product?
- We already know the answer to this question
- It is the Dirac distribution!

$$x = x * \delta = \delta * x$$

• The convolution product is also *associative*, that is, for three signals u, v, and w, we have (check this yourself)

$$(u * v) * w = u * (v * w)$$

- This property can be exploited to determine the total impulse function of two LTI systems interconnected in cascade
- System 1: input signal x(t), impulse function  $h_1(t)$ , output signal u(t)
- System 2: input signal u(t), impulse function  $h_2(t)$ , output signal y(t)
- We assume that System 2 does not "load" System 1



• Response of the total system:

$$y = u * h_2 = (x * h_1) * h_2 = x * (h_1 * h_2) = x * h$$

where we have introduced the impulse function of the total system as

$$h = h_1 * h_2 = h_2 * h_1$$

• Note that since the convolution product of two signals commute, we can interchange the order of the subsystems without affecting the output signal y(t) (provided both systems do not "load" each other)



- Computing the convolution integral: Example 1
- We compute the convolution of the rectangular pulse p with itself
- Recall that

$$p(t) = \begin{cases} 1 & \text{for } 0 < t < 1 \\ 0 & \text{otherwise} \end{cases}$$

• By definition, we have

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

• Since p vanishes outside the interval (0,1), the integral simplifies to

$$y(t) = \int_{\tau=0}^{1} p(t-\tau) d\tau$$

- It is convenient to rewrite this integral in such a way that the time coordinate *t* appears in the integration limits of the integral
- We use the substitution  $t' = t \tau$  to achieve this and arrive at

$$y(t) = \int_{t'=t-1}^{t} p(t') dt'$$

• Now observe that for t < 0 we integrate over an interval outside the support of p. Consequently,

$$y(t) = 0$$
 for  $t < 0$ 

• Similarly, for t-1>1 we again integrate over an interval outside the support of p. We have

$$y(t) = 0$$
 for  $t > 2$ 

• For 0 < t < 1 the lower bound falls outside of the support of p, while the upper bound belongs to this support. We have

$$y(t) = \underbrace{\int_{t'=t-1}^{0} p(t') dt'}_{=0} + \int_{t'=0}^{t} p(t') dt' = \int_{t'=0}^{t} dt' = t \quad \text{for } 0 < t < 1$$

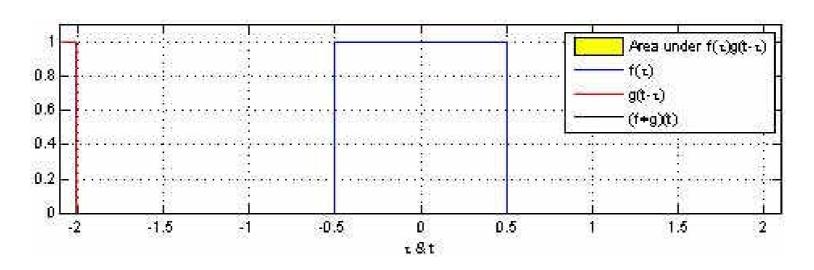
• Finally, for 1 < t < 2 the upper bound falls outside of the support of p and the lower bound is in the support of p. In this case, we have

$$y(t) = \int_{t'=t-1}^{1} p(t') dt' + \underbrace{\int_{t'=1}^{t} p(t') dt'}_{=0} = \int_{t'=t-1}^{1} dt' = 2 - t \quad \text{for } 1 < t < 2$$

• Putting everything together, we find

$$y(t) = \Lambda(t)$$

• The convolution of the rectangular pulse with itself produces the triangular pulse function



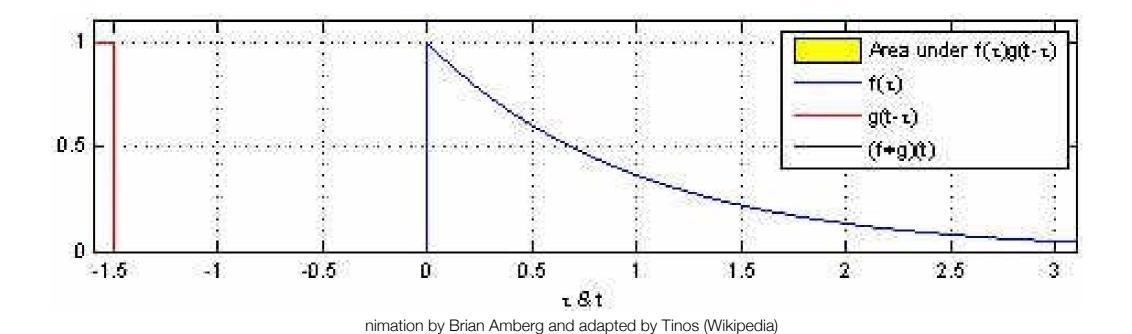
Animation by Brian Amberg and adapted by Tinos (Wikipedia)

- Computing the convolution integral: Example 2
- We graphically determine the convolution of p(t) and p(t/10) (we use the blackboard for this)
- Finally, if the support of a signal x is  $(\ell_x, u_x)$  and the support of a signal h is  $(\ell_h, u_h)$  then

the support of y(t) = x(t) \* h(t) is  $(\ell_x + \ell_h, u_x + u_h)$ 

Verify this statement!

### Another example:



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- Up till now we have been looking at fairly general systems whose action on the input signal is described by some operator *S*
- Let us now be more specific and consider systems described by the linear ordinary differential equation

$$\left(a_N \frac{d^N}{dt^N} + a_{N-1} \frac{d^{N-1}}{dt^{N-1}} + \dots + a_1 \frac{d}{dt} + a_0\right) y(t) = \\
\left(b_M \frac{d^M}{dt^M} + b_{M-1} \frac{d^{M-1}}{dt^{M-1}} + \dots + b_1 \frac{d}{dt} + b_0\right) x(t)$$

which holds for t > 0

• N and M are positive integers

- x(t) is the prescribed input signal
- y(t) is the desired output signal
- To obtain the output signal y(t), we also need the N initial conditions

$$y(0)$$
 and  $\frac{d^k y(t)}{dt^k}\Big|_{t=0}$  for  $k = 1, 2, ..., N-1$ 

• RLC circuits, mechanical systems, etc. can all be described by a differential equation of the above form

- Further on we will show you how to solve the differential equation using the Laplace transform
- For now it suffices to say that the solution y(t) is given by

$$y(t) = y_{\rm ZS}(t) + y_{\rm Zi}(t)$$

- $y_{zs}(t)$  is called the *zero-state* response. This is the solution exclusively due to the input with the initial conditions set to zero
- $y_{zi}(t)$  is called the *zero-input* response. This is the solution exclusively due to the initial conditions with the input set to zero

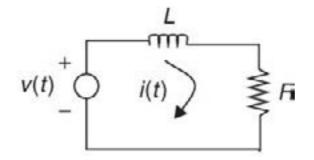
- For vanishing initial conditions the system is linear and time-invariant (LTI). This can easily be seen from the differential equation (check this for yourself)
- In this case the zero-input response vanishes and the solution is equal to the zero-state response
- For nonvanishing initial conditions, the system is no longer an LTI system



- **Example.** Consider a circuit consisting of a resistor of R in series with an inductor L and a voltage source v(t) = Bu(t). The initial current in the inductor is  $I_0$ . The input signal of the system is v(t), the current i(t) in the circuit is the output signal.
- From Kirchhoff's voltage law:

$$L\frac{\mathrm{d}i(t)}{\mathrm{d}t} + Ri(t) = v(t) \quad t > 0$$

with initial condition  $i(0) = I_0$ 



• The output signal is given by

$$i(t) = i_{zs}(t) + i_{zi}(t), t > 0$$

• with

$$i_{\text{ZS}}(t) = \frac{B}{R}(1 - e^{-t/\tau}), \quad i_{\text{Zi}}(t) = I_0 e^{-t/\tau}, \quad \text{and} \quad \tau = L/R$$

If we double the amplitude of the input signal then the output signal becomes

$$i(t) = 2i_{\rm ZS}(t) + i_{\rm Zi}(t)$$

with  $i_{zs}(t)$  and  $i_{zi}(t)$  as above

- We observe that the output is *not* doubled, since  $i_{zi}(t)$  does not vanish
- However, for  $I_0 = 0$  (vanishing initial condition) we do have  $i_{zi}(t) = 0$  and the output is doubled in this case
- With vanishing initial conditions, the system is linear

- A continuous-time system *S* is *causal* if
  - \* whenever its input x(t) = 0, and there are no initial conditions, the output is y(t) = 0
  - \* the output y(t) does not depend on future inputs

• An LTI system is causal if

$$h(t) = 0$$
 for  $t < 0$  (causal LTI system)

• Indeed, for an LTI system we have the convolution integral

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

and writing this integral as

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



- we observe that in the second integral integration takes place over future inputs
- For a causal LTI system, these inputs cannot contribute to the output signal at time instant *t*
- Consequently, for a causal system we must have  $h(t-\tau)=0$  for  $t<\tau<\infty$  or h(t)=0 for t<0
- In case the LTI system is causal we are left with

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

• In addition, if the input signal also vanishes prior to t = 0, that is, if x(t) = 0 for t < 0, then the convolution integral simplifies even further. In this case we have

$$y(t) = \int_{\tau=0}^{t} x(\tau)h(t-\tau) d\tau$$

- Finally, we discuss the concept of BIBO stability
- BIBO stands for **B**ounded **I**nput **B**ounded **O**utput
- We are given a bounded input signal x(t), that is, a signal that satisfies

$$|x(t)| \le M$$

for some positive *M* 

• We ask: Under what condition(s) is the output signal y(t) also bounded?

• To answer this question, consider

$$|y(t)| = \left| \int_{\tau = -\infty}^{\infty} x(t - \tau) h(\tau) \, d\tau \right|$$

$$\leq \int_{\tau = -\infty}^{\infty} |x(t - \tau)| |h(\tau)| \, d\tau$$

$$\leq M \int_{\tau = -\infty}^{\infty} |h(\tau)| \, d\tau$$

• From this inequality it follows that if

$$\int_{\tau=-\infty}^{\infty} |h(\tau)| \, \mathrm{d}\tau < \infty$$

then the output signal y(t) is bounded as well

- If the impulse response is absolutely integrable (the action of the impulse response is finite) then the output is bounded as well
- An LTI system is called *BIBO stable* if the impulse response is absolutely integrable