

Systems Laboratory, Spring 2025

Damiano Varagnolo – CC-BY-4.0

1D convolution in continuous time

Contents map

<u>developed content units</u>	<u>taxonomy levels</u>
convolution	u1, e1

<u>prerequisite content units</u>	<u>taxonomy levels</u>
signal	u1, e1

Main ILO of sub-module “1D convolution in continuous time”

Compute the convolution between two single dimensional continuous time signals

Why convolution?

because for a LTI system with impulse response $h(t)$
it follows that $y_{\text{forced}}(t) = u * h(t)$

extremely important result for LTI systems:

$$y_{\text{forced}}(t) = h * u(t) = u * h(t) :=$$

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

... and this module = what that formula actually means from graphical perspectives

Additional material

Videos:

- <https://www.youtube.com/watch?v=KuXjwB4LzSA>
- <https://www.youtube.com/watch?v=acAw5WGtzuk>
- <https://www.youtube.com/watch?v=IaSGqQa50-M> (for connections with probability)
- https://www.youtube.com/playlist?list=PL4iThgVpN7hmbIhHnCa7SD00gLMoNwED_
- <https://www.youtube.com/playlist?list=PL4mJLdGEHNvhCuPXsKFrnD7AaQB1MEB6a>

Animations:

- <https://lpsa.swarthmore.edu/Convolution/CI.html>
- <https://phiresky.github.io/convolution-demo/>

Towards decomposing this formula in pieces

$$y_{\text{forced}}(t) = h * u(t) = \int_{-\infty}^{+\infty} u(\tau)h(t - \tau)d\tau = \int_{-\infty}^{+\infty} h(\tau)u(t - \tau)d\tau$$

Towards decomposing this formula in pieces

$$y_{\text{forced}}(t) = h * u(t) = \int_{-\infty}^{+\infty} u(\tau)h(t - \tau)d\tau = \int_{-\infty}^{+\infty} h(\tau)u(t - \tau)d\tau$$

better focus on

$$\int_{-\infty}^{+\infty} u(\tau)h(t - \tau)d\tau$$

or on

$$\int_{-\infty}^{+\infty} h(\tau)u(t - \tau)d\tau ?$$

Towards decomposing this formula in pieces

$$y_{\text{forced}}(t) = h * u(t) = \int_{-\infty}^{+\infty} u(\tau)h(t - \tau)d\tau = \int_{-\infty}^{+\infty} h(\tau)u(t - \tau)d\tau$$

better focus on

$$\int_{-\infty}^{+\infty} u(\tau)h(t - \tau)d\tau$$

or on

$$\int_{-\infty}^{+\infty} h(\tau)u(t - \tau)d\tau ?$$

in automatic control typically better the second

Towards decomposing this formula in pieces, small change of notation

$$y_{\text{forced}}(t) = \int_{-\infty}^{+\infty} h(\tau) u(t - \tau) d\tau \quad \mapsto \quad y_{\text{forced}}(\text{now}) = \int_{-\infty}^{+\infty} h(\tau) u(\text{now} - \tau) d\tau$$

Decomposing this formula in pieces

$$y_{\text{forced}}(\text{now}) = \int_{-\infty}^{+\infty} h(\tau) u(\text{now} - \tau) d\tau$$

\implies constituent pieces =

- $u(\text{now} - \tau)$
- $h(\tau)$
- $u(\text{now} - \tau)h(\tau)$
- $\int u(\text{now} - \tau)h(\tau) d\tau$

Visualizing the various pieces

$$u(\text{now} - \tau) \quad h(\tau) \quad u(\text{now} - \tau)h(\tau) \quad \int_{-\infty}^{+\infty} h(\tau)u(\text{now} - \tau)d\tau$$

time

τ

Example

$$u(t) = \begin{cases} 1 & \text{for } t \in [1, 2] \\ 0 & \text{otherwise} \end{cases}$$

$$h(t) = \begin{cases} 2 & \text{for } t \in [0, 1) \\ 1 & \text{for } t \in [1, 2] \\ 0 & \text{otherwise} \end{cases}$$

signals(t)



signals(τ)



Another Example:

$$u(t) = \begin{cases} 1 & \text{for } t \in [1, 2] \text{ and } t \in [2, 3] \\ 0 & \text{otherwise} \end{cases}$$

$$h(t) = \begin{cases} 2 & \text{for } t \in [0, 2) \\ 0 & \text{otherwise} \end{cases} \quad ?$$

Paramount message

h in $y_{\text{forced}}(t) = \int_{-\infty}^{+\infty} h(\tau) u(t - \tau) d\tau$ represents how much the past u 's contribute to the current y_{forced} :



Refreshing what we are doing and why

Dynamics of a cart: $\dot{v}(t) = -\frac{k}{m}v(t) + \frac{k}{m}F(t)$ with:

- **control input:** $u(t)$ (actuation from the motor, in this case $= F(t)$)

Refreshing what we are doing and why

Dynamics of a cart: $\dot{v}(t) = -\frac{k}{m}v(t) + \frac{k}{m}F(t)$ with:

- **control input:** $u(t)$ (actuation from the motor, in this case = $F(t)$)
- **system output:** $y(t)$ (cart velocity, in this case = $v(t)$)

Refreshing what we are doing and why

Dynamics of a cart: $\dot{v}(t) = -\frac{k}{m}v(t) + \frac{k}{m}F(t)$ with:

- **control input:** $u(t)$ (actuation from the motor, in this case = $F(t)$)
- **system output:** $y(t)$ (cart velocity, in this case = $v(t)$)
- **impulse response:** $h(t)$ (output corresponding to the input $\delta(t)$ assuming $y(0) = 0$)

Refreshing what we are doing and why

Dynamics of a cart: $\dot{v}(t) = -\frac{k}{m}v(t) + \frac{k}{m}F(t)$ with:

- **control input:** $u(t)$ (actuation from the motor, in this case = $F(t)$)
- **system output:** $y(t)$ (cart velocity, in this case = $v(t)$)
- **impulse response:** $h(t)$ (output corresponding to the input $\delta(t)$ assuming $y(0) = 0$)
- **free evolution:** $y_{\text{free}}(t)$ (output in time corresponding to no input, i.e., $u(t) = 0$, and initial condition $y(0)$ whatever it is)

Refreshing what we are doing and why

Dynamics of a cart: $\dot{v}(t) = -\frac{k}{m}v(t) + \frac{k}{m}F(t)$ with:

- **control input:** $u(t)$ (actuation from the motor, in this case $= F(t)$)
- **system output:** $y(t)$ (cart velocity, in this case $= v(t)$)
- **impulse response:** $h(t)$ (output corresponding to the input $\delta(t)$ assuming $y(0) = 0$)
- **free evolution:** $y_{\text{free}}(t)$ (output in time corresponding to no input, i.e., $u(t) = 0$, and initial condition $y(0)$ whatever it is)
- **forced response:** $y_{\text{forced}}(t) = u * h(t)$ (output in time corresponding to null initial condition, i.e., $y(0) = 0$, and input $u(t)$ whatever it is)

Refreshing what we are doing and why

Dynamics of a cart: $\dot{v}(t) = -\frac{k}{m}v(t) + \frac{k}{m}F(t)$ with:

- **control input:** $u(t)$ (actuation from the motor, in this case $= F(t)$)
- **system output:** $y(t)$ (cart velocity, in this case $= v(t)$)
- **impulse response:** $h(t)$ (output corresponding to the input $\delta(t)$ assuming $y(0) = 0$)
- **free evolution:** $y_{\text{free}}(t)$ (output in time corresponding to no input, i.e., $u(t) = 0$, and initial condition $y(0)$ whatever it is)
- **forced response:** $y_{\text{forced}}(t) = u * h(t)$ (output in time corresponding to null initial condition, i.e., $y(0) = 0$, and input $u(t)$ whatever it is)
- **total response:** $y(t) = y_{\text{free}}(t) + y_{\text{forced}}(t)$

Quiz time!

$$h * u(t) := \int_{-\infty}^{+\infty} h(\tau) u(t - \tau) d\tau$$

- is $h * u(t) = u * h(t)$?

Quiz time!

$$h * u(t) := \int_{-\infty}^{+\infty} h(\tau) u(t - \tau) d\tau$$

- is $h * u(t) = u * h(t)$?
- is $(\alpha h_1 + \beta h_2) * u(t) = \alpha (h_1 * u(t)) + \beta (h_2 * u(t))$?

Quiz time!

$$h * u(t) := \int_{-\infty}^{+\infty} h(\tau) u(t - \tau) d\tau$$

- is $h * u(t) = u * h(t)$?
- is $(\alpha h_1 + \beta h_2) * u(t) = \alpha (h_1 * u(t)) + \beta (h_2 * u(t))$?
- if both $h(\tau) = 0$ and $u(t) = 0$ if $t < 0$, how can we simplify $y(t) = \int_{-\infty}^{+\infty} h(\tau) u(t - \tau) d\tau$?

Summarizing

Compute the convolution between two single dimensional continuous time signals

- take one of the two signals
- translate it to the “current t ”
- flip it
- multiply the two signals in a pointwise fashion
- compute the integral of the result

Most important python code for this sub-module

Methods implementing (discrete) convolutions

- <https://numpy.org/doc/2.1/reference/generated/numpy.convolve.html>
- <https://docs.scipy.org/doc/scipy/reference/generated/scipy.signal.convolve.html>

Self-assessment material

Question 1

What does the convolution integral $y_{\text{forced}}(t) = \int_{-\infty}^{+\infty} h(\tau)u(t-\tau)d\tau$ represent in the context of LTI systems?

Potential answers:

- I: The free evolution of the system output.
- II: The forced response of the system output due to the input $u(t)$.
- III: The total response of the system, including initial conditions.
- IV: The impulse response of the system.
- V: I do not know.

Question 2

Which of the following is true about the convolution operation $h * u(t)$?

Potential answers:

- I: It is only defined for periodic signals.
- II: It is only applicable to discrete-time systems.
- III: It is commutative, i.e., $h * u(t) = u * h(t)$.
- IV: It requires both signals to be symmetric.
- V: I do not know.

Question 3

What does the impulse response $h(t)$ of an LTI system represent?

Potential answers:

- I: The input signal $u(t)$ applied to the system.
- II: The free evolution of the system output.
- III: The total response of the system, including initial conditions.
- IV: The output of the system when the input is a Dirac delta function $\delta(t)$.
- V: I do not know.

Question 4

If $h(\tau) = 0$ for $\tau < 0$ and $u(t) = 0$ for $t < 0$, how can the convolution integral $y(t) = \int_{-\infty}^{+\infty} h(\tau)u(t - \tau)d\tau$ be simplified?

Potential answers:

I: $y(t) = \int_0^t h(\tau)u(t - \tau)d\tau$

II: $y(t) = \int_0^{+\infty} h(\tau)u(t - \tau)d\tau$

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

IV: $y(t) = \int_{-\infty}^0 h(\tau)u(t - \tau)d\tau$

V: I do not know.

Question 5

What is the graphical interpretation of $h(\tau)$ in the convolution integral

$$y_{\text{forced}}(t) = \int_{-\infty}^{+\infty} h(\tau) u(t - \tau) d\tau?$$

Potential answers:

- I: It represents the future inputs of the system.
- II: It represents how much past inputs contribute to the current output.
- III: It represents the free evolution of the system.
- IV: It represents the total energy of the system.
- V: I do not know.

Recap of sub-module “1D convolution in continuous time”

- convolution is an essential operator, since it can be used for LTI systems to compute forced responses
- its graphical interpretation aids interpreting impulse responses as how the past inputs contribute to current outputs

?