

# Linear Shift Invariant Systems

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# Linear Shift Invariant Systems

- 1 Image Processing (Chapter 3 from Szeliski and Chapter 6 from Horn)

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# Continuous Domain

- Key task in Image Processing: Transform an image to a form more amenable to further manipulation.
- Imaging systems or image formation systems can be approximated by Linear Shift Invariant Systems (LSI), a powerful analytic tool.
- Consider a 2D system with inputs  $f_1(x, y)$  and  $f_2(x, y)$  and outputs  $g_1(x, y)$  and  $g_2(x, y)$ .

A set of small, light-blue navigation icons typically found in Beamer presentations, including symbols for back, forward, search, and other navigation functions.

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- System is linear if  $Input = \alpha f_1 + \beta f_2$  and  $Output = \alpha g_1 + \beta g_2$ .
- Which of the following systems are linear? (a)  $g(x) = e^x f(x)$ , (b)  $g(x) = f(x) + 1$ , (c)  $g(x) = xf(x)$ , (d)  $g(x) = (f(x))^2$
- **Shift Invariant:**  $Input = f(x - a, y - b)$ ,  $Output = g(x - a, y - b)$ , for arbitrary  $a, b$ . In practice, images are limited in area and so, shift invariance holds only for limited areas.

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# Why are LSIs important?

- Can model image formation systems using LSIs
- System shortcomings can be discussed in terms of the LSI model.
- Study of LSI leads to useful algorithms for processing images digitally or even optically.

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# Convolution and Point Spread Functions

- Consider a system with  $f(x, y)$  as input that produces,

$$g(x, y) = \iint_{-\infty}^{\infty} f(x - \xi, y - \eta) h(\xi, \eta) d\xi d\eta \quad (1)$$

$$\text{i.e. } g_{out} = f_{in} \otimes h_{in} \quad (\text{convolution integral}) \quad (2)$$

- This system is an LSI system:

$$\text{Input : } \alpha f_1 + \beta f_2; \text{ Output : } \alpha g_1 + \beta g_2$$

$$\text{Input : } f(x - a, y - b); \text{ Output : } g(x - a, y - b)$$

- Thus, *a system whose response is described by a convolution,  $\otimes$  is an LSI and conversely, any LSI system performs a convolution.*


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# Convolution Example

## Review of convolution

- Illustration of  $h(x) = f(x) * g(x) = \int_{-\infty}^{\infty} f(u)g(x-u)du$

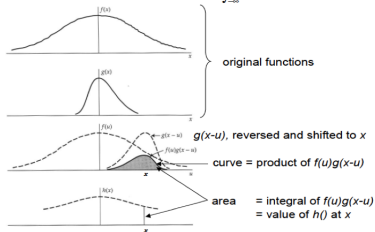


Figure: Convolution Example

# Impulse Functions

- **Question:** Given an arbitrary function  $h(x, y)$ , can we find a function  $f(x, y)$  that causes the output to be  $h(x, y)$  i.e.,

$$h(x, y) = \iint_{-\infty}^{\infty} f(x - \xi, y - \eta) h(\xi, \eta) d\xi d\eta \quad (3)$$

- **Ans:** Yes; called an *impulse function*.

$$\begin{aligned} \delta(x, y) &= 0 \text{ if } (x, y) \neq 0 \\ \iint_{-\infty}^{\infty} \delta(x, y) &= 1 \end{aligned}$$

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# Impulse Function Definition & Properties

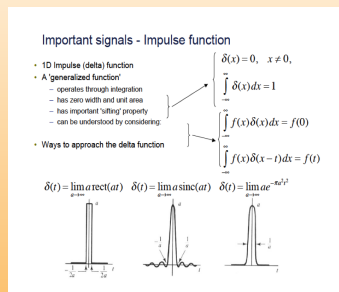


Figure: Impulse Function

- Further,  $\delta(x, y) \otimes f(x, y) = f(x, y) \otimes \delta(x, y) = f(x, y)$ .
- $\delta(ax, by) = (1/|ab|)\delta(x, y)$

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# Sampling and Replicating Property

- Consider an infinite sequence of impulses denoted by the “comb” or “shah” symbol,

$$\text{III}(x, y) = \sum_{-\infty}^{\infty} \sum_{-\infty}^{\infty} \delta(x - n, y - m) \quad (4)$$

- Properties:

$$\text{III}(ax, by) = \frac{1}{|ab|} \sum_{-\infty}^{\infty} \sum_{-\infty}^{\infty} \delta(x - n/a, y - m/b) \quad (5)$$

$$\text{III}(-x, -y) = \text{III}(x, y) \quad (6)$$

$$\iint_{-1/2}^{1/2} \text{III}(x, y) dx dy = 1 \quad \text{because its periodic with unit period} \quad (7)$$

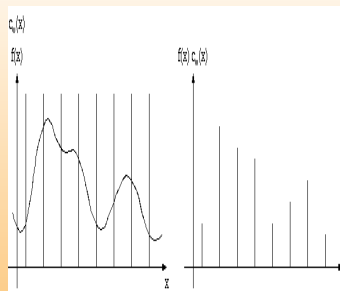


Figure: Sampling a function

- Multiplication of a function by  $\text{III}(x, y)$  effectively samples it at unit intervals.

$$\text{III}(x, y)f(x, y) = \sum_{n=-\infty}^{\infty} \sum_{m=-\infty}^{\infty} f(n, m)\delta(x - n, y - m) \quad (8)$$



$$\text{III}(x, y) \otimes f(x, y) = \sum_{-\infty}^{\infty} \sum_{-\infty}^{\infty} f(x - n, y - m)$$

- **Seaparability:**  $\delta(x, y) = \delta(x)\delta(y)$ , similarly,  $\text{III}(x, y) = \text{III}(x)\text{III}(y)$ .
- **Derivative of the Impulse function:** We use the limit definition of the impulse function and take the derivative of the function whose limit defines the impulse function. Then take the limit.
- **Derivative Sifting Property:**

$$\delta'(x, y) \otimes f(x, y) = \iint_{-\infty}^{\infty} \delta'(x - \xi, y - \eta) f(\xi, \eta) d\xi d\eta = f'(x, y)$$

- Similarly,  $\delta''(x, y) \otimes f(x, y) = f''(x, y)$

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# LSI Performs Convolution

- **Proof:** Using the sifting property of the impulse function,

$$f(x, y) = \iint_{-\infty}^{\infty} f(\xi, \eta) \delta(x - \xi, y - \eta) d\xi d\eta \quad (9)$$

- Decompose  $f(x, y)$  into elementary functions and then we can determine the overall output  $g(x, y)$  by summing the shifted scaled impulses. Why? (because we can use the fact that the system is linear)
- Response of the LSI system to  $\alpha \delta(x - \xi, y - \eta)$  is  $\alpha h(x - \xi, y - \eta)$  as the system is shift invariant. But  $\alpha = f(\xi, \eta)$ .
- Hence,

$$g(x, y) = \iint_{-\infty}^{\infty} f(\xi, \eta) h(x - \xi, y - \eta) d\xi d\eta \quad (10)$$

Which is the same as  $g(x, y) = f(x, y) \otimes h(x, y)$ . QED.

# Properties of Convolution

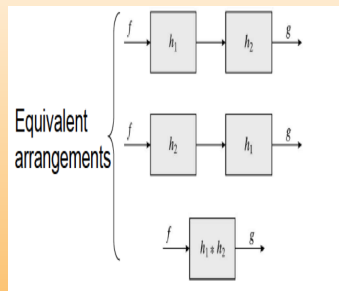


Figure: Convolution Properties

- Convolution is commutative:  $f \otimes g = g \otimes f$ .
- Convolution is associative:  $(f \otimes g) \otimes h = f \otimes (g \otimes h)$ .
- Can cascade systems etc.

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