

Convolution

Overview

Class Meeting : 20 Jan 2026

- convolution
 - analytical
 - discrete
 - numerical
- formulate in terms of convolution kernels
- compute convolutions in one and higher dimension numerically
- Gaussian convolution
- box convolution
- efficiency and benefit of separable kernels

Convolution

[optional]

- convolution is a mathematical **technique used** on two **functions f and g** that **produces** a third **function $f * g$**
- $f * g$ is the **integral** of the **product** of the two **functions** after **one** is **reflected** about the **y-axis** and **shifted**
- uses an **average** of **local-adjacent values** based on kernel size

<https://en.wikipedia.org/wiki/Convolution>

Convolution in 1D

Applications of Convolution

- signal processing
 - filtering (noise reduction)
 - system modeling
 - reverberation (simulating sound reflections)
- image processing
 - sharpening
 - edge detection
 - feature extraction
- artificial intelligence

	<ul style="list-style-type: none"> ○ convolutional neural networks ● physics and engineering <p>https://www.fieldbox.ai/seeing-through-computer-vision-convolution-101/</p>
General Mathematical form	$(f * g)(t) = \int_{-\infty}^{\infty} f(\tau)g(t - \tau)d\tau = \int_{-\infty}^{\infty} f(t - \tau)g(\tau)d\tau$
Properties	<ul style="list-style-type: none"> ● $f * g = g * f$ ● $f * (g * f) = (f * g) * h$ ● $f * (g + h) = (f * g) + (f * h)$ <p>https://mathworld.wolfram.com/Convolution.html</p>
Discrete Convolution	$(f * g)(i) = \sum_k f(k)g(i - k) = \sum_k f(i - k)g(k)$
Layering	<ul style="list-style-type: none"> ● a weighted sum of the convolution kernel ● can extract local features
Convolution Kernel	<ul style="list-style-type: none"> ● the filter of a matrix for feature extraction ● each convolution process has its own kernel ● image blurring example: for each pixel in an image, the inner product of the pixel within the local window centered on that pixel and the kernel is calculated ● some kernel types <ul style="list-style-type: none"> ○ asymmetric ○ hat ○ box ○ exponential decay
Padding	<ul style="list-style-type: none"> ● for when the convolution range is not defined at edges

- **f** is padded with reasonable **data** there are **multiple tactics** available
 - **zeros**
 - **mirror-copy**
 - **repeat closest value**
 - **trim the range** (usually only with **very large sets**)

Convolution in 2D

Overview	<ul style="list-style-type: none"> • two-dimensional convolution is mathematically the same as in 1D • produced 2D integral • often kernels are separable
Kernel Separability	<ul style="list-style-type: none"> • a function $g(x, y)$, that can be rewritten as a product such that: $g(x, y) = g^x(x) * g^y(y)$ • the result is two 1D kernels • by separating the kernel, each dimension can be convoluted individually • continuous separable $(f * g)(x, y) = \int_{-R}^R \int_{-R}^R f(x - u, y - v)g(u, v)dudv$ • discrete separable $(f * g)(i, j) = \sum_{I=-k}^k \sum_{J=-k}^k f(i - I, j - J)g(I, J)$
Sources	<ul style="list-style-type: none"> • https://mathworld.wolfram.com/Convolution.html • https://evidentscientific.com/en/microscope-resource/tutorials/digital-imaging/processing/convolutionkernels

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Live Session Notes		30 Oct 2025
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