

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

	<ul style="list-style-type: none"> <li>• <b>f</b> is <b>padding</b> with reasonable <b>data</b> there are <b>multiple tactics</b> available <ul style="list-style-type: none"> <li>○ <b>zeros</b></li> <li>○ <b>mirror-copy</b></li> <li>○ <b>repeat closest value</b></li> <li>○ <b>trim the range</b> (usually only with <b>very large sets</b>)</li> </ul> </li> </ul>
<b>Convolution in 2D</b>	
<b>Overview</b>	<ul style="list-style-type: none"> <li>• two-dimensional convolution is mathematically the same as in 1D</li> <li>• produced 2D integral</li> <li>• often kernels are separable</li> </ul>
<b>Kernel Separability</b>	<ul style="list-style-type: none"> <li>• a function <math>g(x, y)</math>, that can be rewritten as a product such that:  <math>g(x, y) = g^x(x) * g^y(y)</math></li> <li>• the result is two 1D kernels</li> <li>• by separating the kernel, each dimension can be convoluted individually</li> <li>• continuous separable <math display="block">(f * g)(x, y) = \int_{-R}^R \int_{-R}^R f(x - u, y - v) g(u, v) du dv</math> </li> <li>• discrete separable <math display="block">(f * g)(i, j) = \sum_{l=-k}^k \sum_{m=-k}^k f(i - l, j - m) g(l, m)</math> </li> </ul>
<b>Sources</b>	<ul style="list-style-type: none"> <li>• <a href="https://mathworld.wolfram.com/Convolution.html">https://mathworld.wolfram.com/Convolution.html</a></li> <li>• <a href="https://evidentscientific.com/en/microscope-resource/tutorials/digital-imaging/processing/convolutionkernels">https://evidentscientific.com/en/microscope-resource/tutorials/digital-imaging/processing/convolutionkernels</a></li> </ul>

<https://evidentscientific.com/en/microscope-resource/tutorials/digital-imaging/processing/convolutionkernels>

Live Session Notes		30 Oct 2025
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