

# *Convolutional Neural Networks*

**INSTRUCTOR: HONGJIE CHEN  
JUNE 21ST 2022**

# *Types of NNs*

- A normal DNN
  - Time expensive
  - Parameter explosion
  - Universal but unnecessary
- Data format
  - Images (2D), Convolutional Neural Networks
  - Time-series (variable length), Recurrent Neural Networks
  - Networks (graph structure), Graph Neural Networks
  - And more?

# Convolution

- A mathematical operation on two functions  $x()$  and  $w()$  that produces a third function  $y()$  that can be viewed as a modified version of one of the original functions  $x()$

$$y(i) = \int_t x(t)w(i - t)dt, \text{ written as } y(i) = (x * w)(i)$$

- Demo: <https://www.youtube.com/watch?v=das6mpj13XQ>
- Smoothing effect (reading: [link](#))

# Discrete Convolution

- Discrete convolution

$$y(i) = \sum_{t=-\infty}^{\infty} x(t)w(i - t)$$

- Two dimensional convolution or even higher dimensions

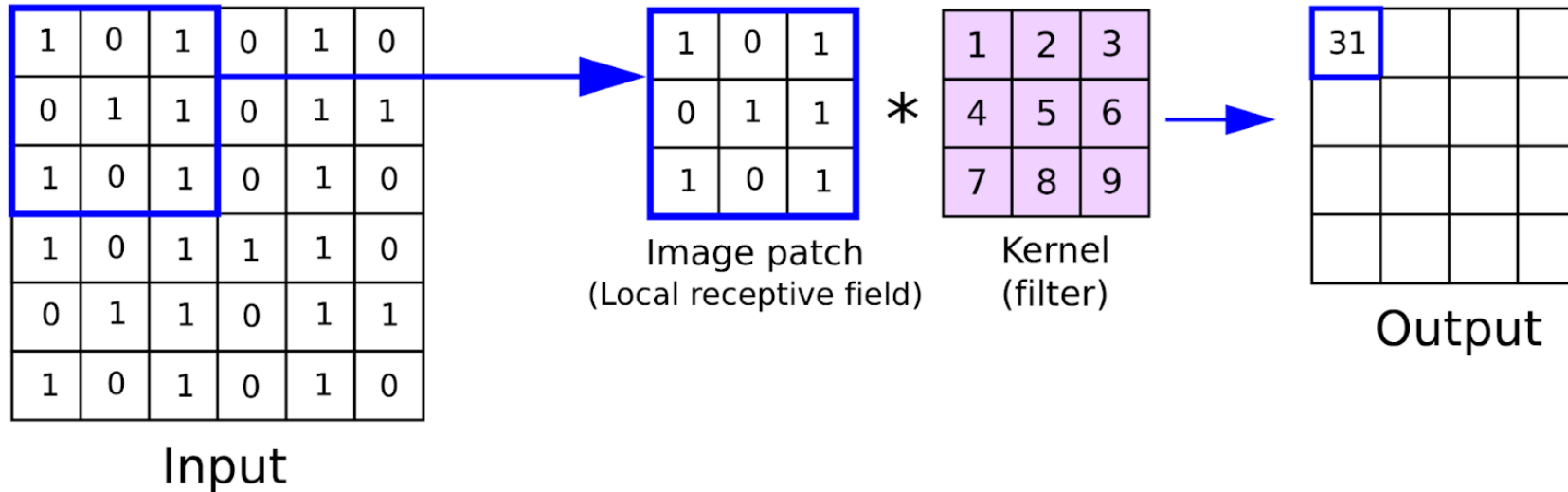
$$y(i, j) = \sum_{t_1=-\infty}^{\infty} \sum_{t_2=-\infty}^{\infty} x(t_1, t_2)w(i - t_1, j - t_2)$$

# *Convolution in Neural Networks*

- A convolution denotes the linear combination of **a subset of units of a specific pattern** of weights
- Instead of all weights as in dense layers

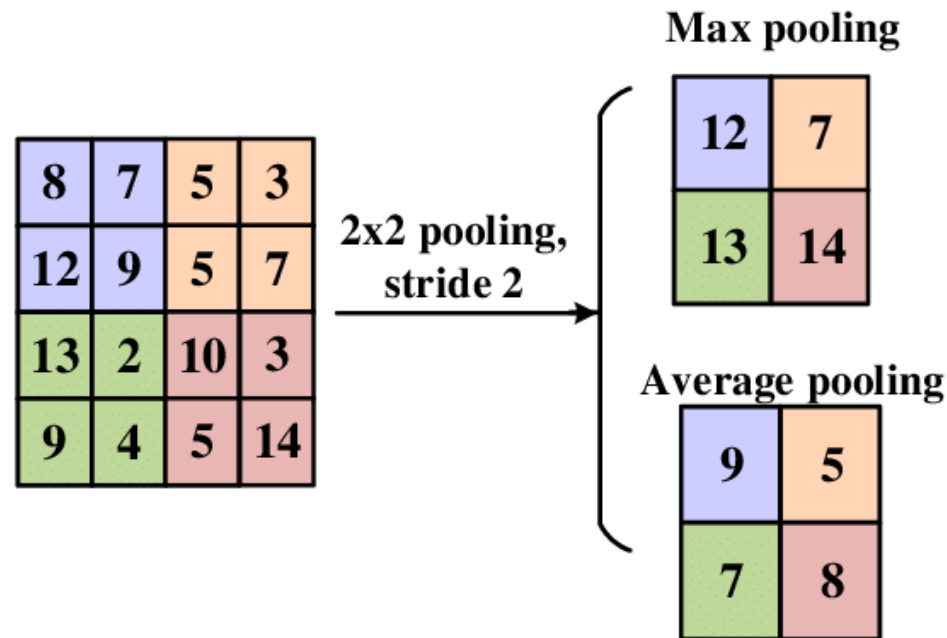
# CNN

- A CNN is a NN that consists of convolutional layers (and optionally pooling layers).



# Pooling

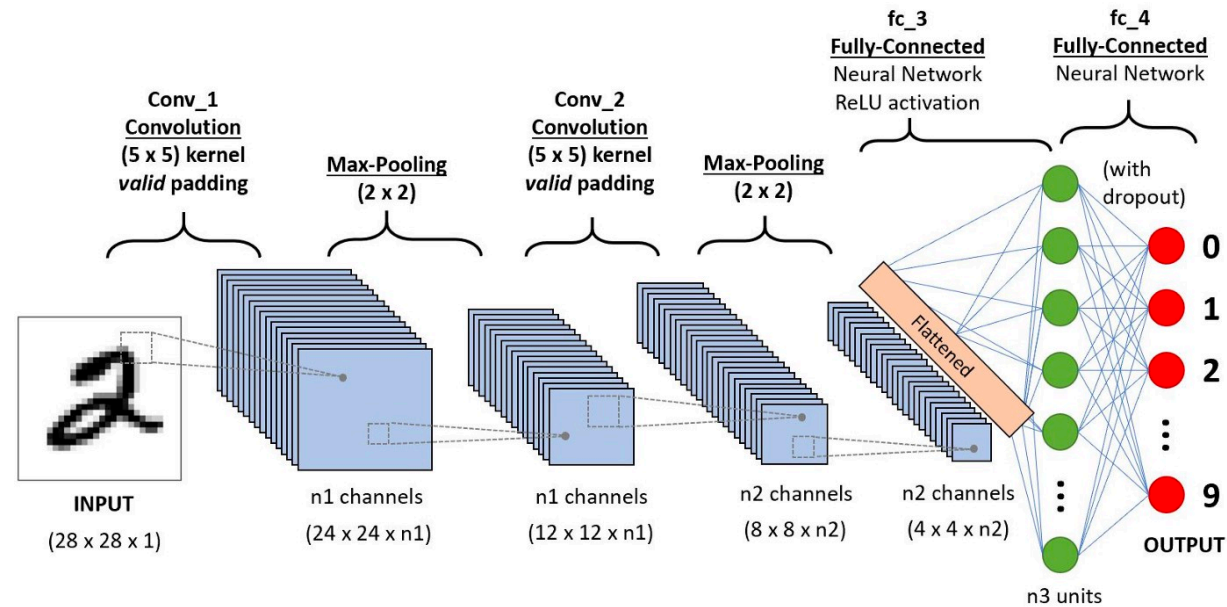
- Downsample features by aggregation



- Popular pooling functions
  - Max, mean, sum, product, etc.

# Classical CNN Example

- Digit recognition



- Back in history: [https://www.youtube.com/watch?v=FwFduRA\\_L6Q](https://www.youtube.com/watch?v=FwFduRA_L6Q)



# Advantages of CNN

- Time: fewer weights to calculate, faster
- Space: fewer links, or equivalently weight parameters
- Locally equivariant representation

- Capture the pattern
- Better Generalization

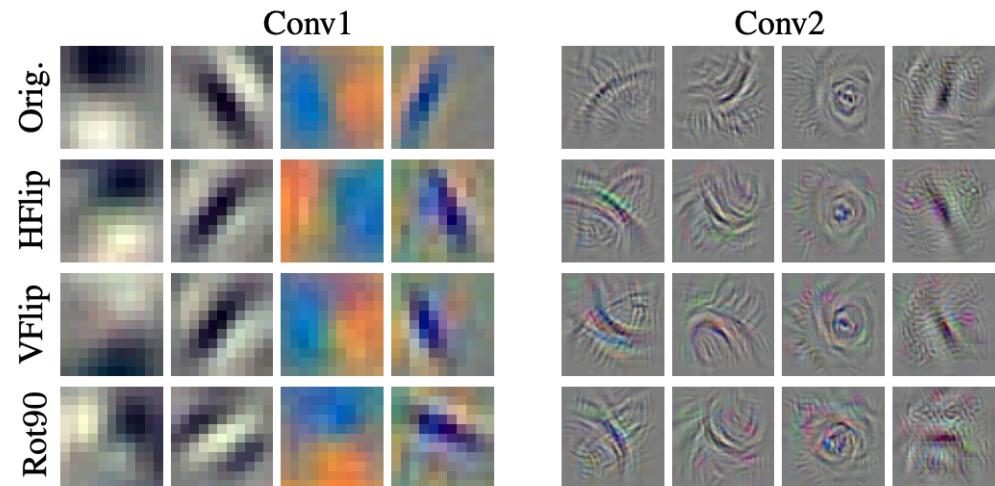


Figure credits: [paper](#)

# Parameters of Convolutional Layers

- **Number of filters:** an integer indicating how many filters / channels for each window where one filter yield one scalar value
- **Kernel size:** a tuple (width, height) indicating the size of the window
- **Stride:** a tuple (horizontal, vertical) indicating shifting step between window
- **Padding:** “valid” or “same” where the former indicates no padding, thus the resulting values will be in a shrinking size, while the latter indicates padding zeros to let the resulting values remain the same size.

\*Don't get confused with the Kernel in kernel perceptron

# *CNN Parameters Example*

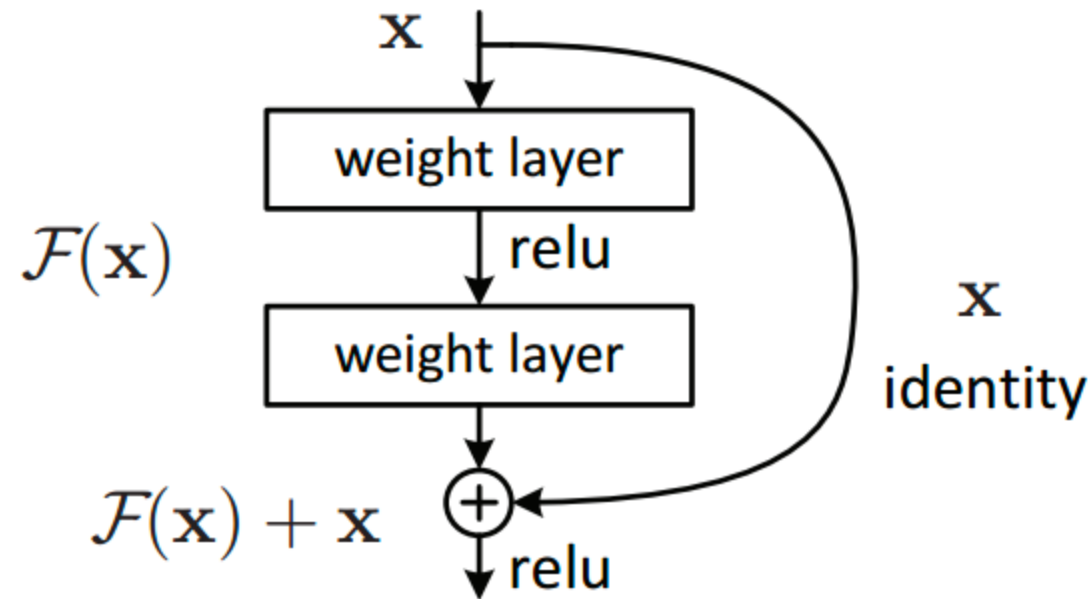
- **Number of filters, Kernel size, Stride, Padding**
- Let's draw a diagram

# *Fewer Large Filters or More Small Filters*

- More small filters
  - Fewer parameters
  - Allow deeper structure

# Residual

- Also called skip connection
  - Address vanishing gradients



# ResNet for Image Recognition (Even deeper)

