# CMSC 373 Artificial Intelligence Fall 2025 16-ConvolutionNetworks

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## **Deep Learning Revolution**

- **Deep Learning** refers to deep neural networks (i.e. many hidden layers)
- The "deep" in Deep learning is NOT "learning that is deep" (i.e. meaningful or sophisticated learning!!)
- The "deep" ONLY refers to the "depth in layers" of the neural network.
- Convolution Networks are a kind of Deep Neural network.

### Taking Inspiration from the Visual Cortex

- Hubel & Wiesel's experiments with cats and primates
  - Visual cortex is a hierarchical series of layers of neurons. Layers communicate back and forth extensively.
  - Layers act as feature detectors (edges, shapes, objects, etc).
  - Each neuron receives input corresponding to a specific small region of the visual scene in a neuron's receptive field).
  - Neurons activate only if their receptive field contains a particular kind of edge/feature (e.g. horizontal edge, vertical edge, angular edge, etc)
  - Lower-level neurons feed into higher level layers of the visual cortex for detecting shapes, objects, faces, etc.
- This is still a gross simplification. The brain is much more complex!

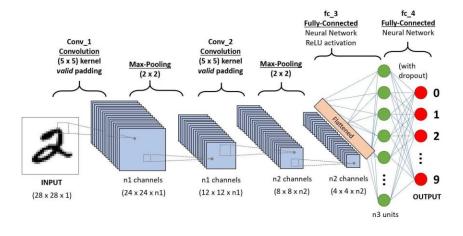


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## Convolution Networks – Short History

- Taking inspiration from Hubel & Wiesel...
- From Cognitron, to NeoCognitron (Fukushima, 1970s)
  Showed how a hierarchical network (using Relu!) could learn using unsupervised means.
- To Convolutional Neural Networks (1989) used for hand-written zipcodes. Developed by Yann LeCun (at AT&T Bell Labs). In 1995, LeCun et al developed LeNet-5 to classify handwritten digits (32x32 pixel images). Used to recognize numbers on checks by banks.

### **ConvNet Architecture**

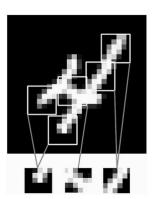


rom: https://saturncloud.io/blog/a-comprehensive-guide-to-convolutional-neural-networks-the-eli5-way

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### Dense Networks versus Convolution Networks

- Layers in dense networks learn global patterns in the input
  - E.g. MINIST Digit recognition: we flattened 28x28 images into 784 units and fed them into the hidden layer.
- Convolution Networks learn local patterns in input
  e.g. They look for patterns in small 2D windows (using patches/filters/kernels) of input images.
- Convolution Networks learn spatial hierarchies of patterns (e.g., edges, larger patterns, etc.)



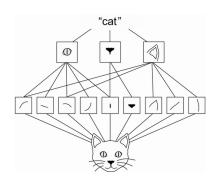
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### Convolution Networks: Core Idea

Convolution Networks learn local patterns in input

e.g. They look for patterns in small 2D windows (using *patches/filters/kernels*) of input images.

 Convolution Networks learn spatial hierarchies of patterns (e.g., edges, larger patterns, etc.)



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#### Convolution Networks: Basic Elements

• Convolution (using filters/patches/kernels)

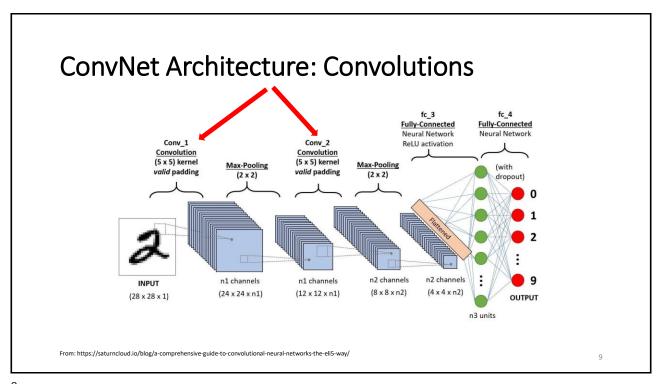
Small patches of input image/map are filtered to recognize local features

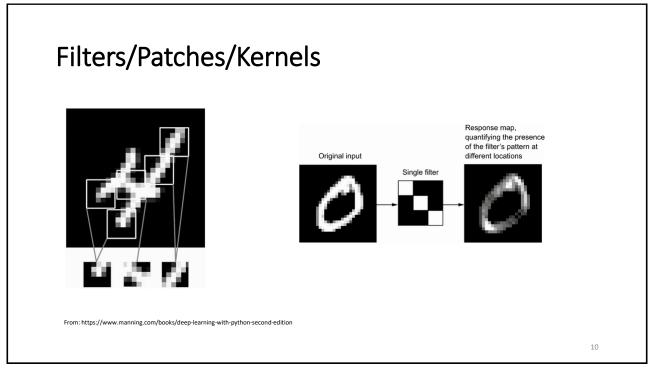
Feature Maps

Starting with an input image (e.g. 28x28x1) Using a 3x3 filter to get K output maps (26x26xK)

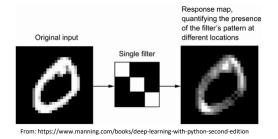
Max Pooling

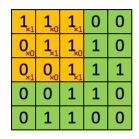
A way of down sampling a feature map.





## Filters/Patches/Kernels







Image

Convolved Feature

From: https://saturncloud.io/blog/a-comprehensive-guide-to-convolutional-neural-networks-the-eli5-way/

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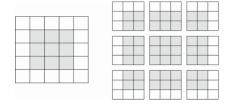
## Filters: Padding

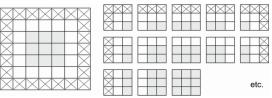
• "valid" Padding

No padding of input is done. Reduces the width and height of resulting map.

• "same" Padding

Pads input in such a way to preserve the width and height of the resulting map.

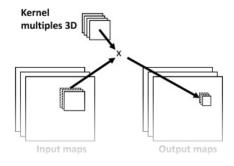


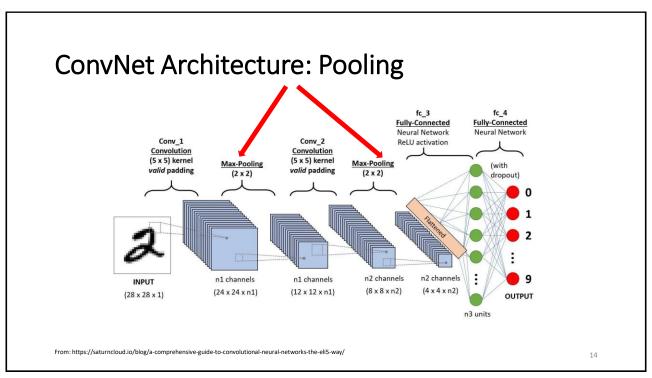


From: https://www.manning.com/books/deep-learning-with-python-second-edition

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## Input and Output Feature Maps





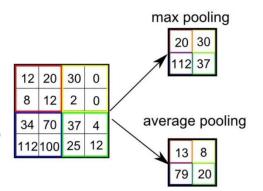
## **Max Pooling**

• Pooling uses a kernel/patch/window

There two popular kinds: **Max pooling**, Average Pooling

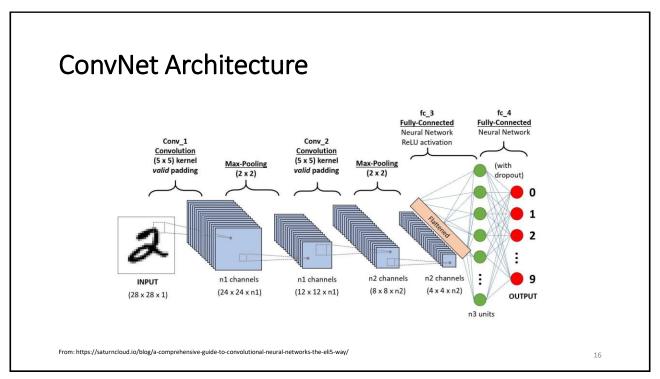
 It is purely an arithmetic operation that helps further downsample a feature map (i.e. performs dimensionality reduction useful for reducing computational load).

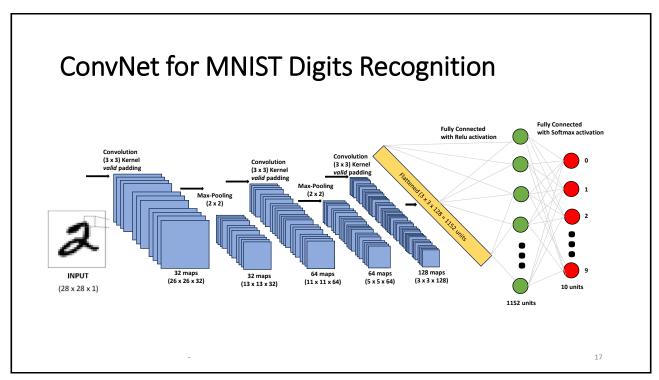
• We typically use a 2x2 window.

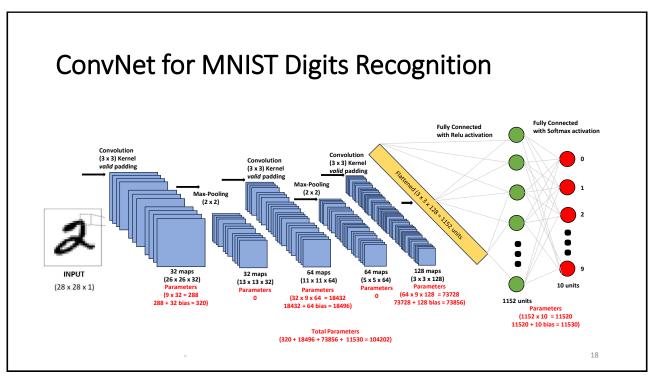


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### Over to Colab

 https://colab.research.google.com/drive/1XqXJiz20EAgHStE2qsPiFa2r kL91uQwp?usp=sharing

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#### ConvNet Architecture Patterns

#### · Overfitting/Dropout

When models tend to overfit, a way to resolve overfitting is to use **dropout**. This is called a **regularization technique**.

A dropout layer can be added in between layers. Typically, a dropout layer eliminates (sets to 0) 20-50% of the outputs (dropout value can be set during model/layer specification).

#### · Data/Batch Normalization

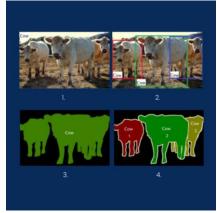
Normalizing the values on inputs and outputs helps with gradient propagation and allows for deeper networks. Hence liberally used in very deep network architectures.

#### Data Augmentation

When datasets are small, "new" data can be created by **transforming** images in the dataset using flipping, rotation, zooming, etc. to augment the dataset with transformed images. Helps avoid overfitting and leads to better generalization.

## **ConvNets Applications**

- Input Image (1)
- Image classification
- Image Segmentation (3, 4)
- Object Detection (3)



From: https://www.smart-interaction.com/2022/07/14/computer-vision-the-ultimate-guide-on-the-4-main-tasks/

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### **CovNet Architectures**

#### AlexNet

Alex Krizhevsky, Ilya SutsKever, Geoffrey Hinton, University of Totonto, 2017.

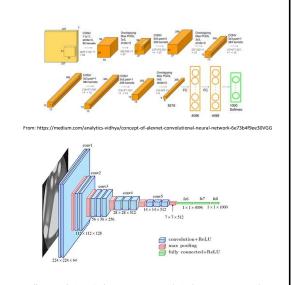
One of the early CNNs to use GPUs to speed up training (in 2012).

#### VGGNet

Karen Simonyan and Andrew Zisserman, University of Oxford, 2014.

Groundbreaking DL model for Object Recognition.

• There are many more.



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## References

- F. Chollet: Deep Learning with Python, Second Edition, Manning2021.
- A. Geron: *Hands-on Machine Learning with Scikit-Learn, Keras, and TensorFlow* 2<sup>nd</sup> Edition. O'Reilly, 2019.
- M. Mitchell: *Artificial Intelligence: A Guide For Thinking Humans,* Farrar, Strouss, Giroux, 2019.
- M. Wooldridge: A Brief History of Artificial Intelligence. Flatiron Books, 2020.