# APS360 Fundamentals of AI

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Lecture 6; Jan 24, 2019

#### Agenda

- Convolutional Neural Networks
- Defining Convolutional Layers

### **Convolutional Neural Networks**

Issue with Fully Connected Layers

- Input image: 200 × 200 pixels
- First hidden layer: 500 units

How many connections are there?

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Answer: 20 million!

Problematic because:

## Issue with Fully Connected Layers

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How many connections are there?

Answer: 20 million!

Problematic because:

- computing predictions takes a long time
- a large number of weights requires a lot of training data to avoid overfitting
- small shift in image can result in large change in prediction

# Convolutional Neural Network

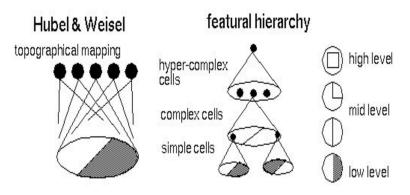
Ideas:

- Locally-connected layers: look for local features in small regions of the image
- Weight-sharing: detect the same local features across the entire image

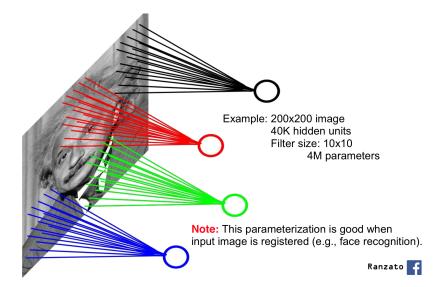
# Convolutional Neural Network

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## Locally Connected Layers



# Weight Sharing

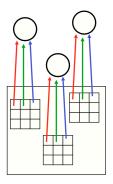
Share the same parameters across different locations (assuming input is stationary): Convolutions with learned kernels



# Weight Sharing (continued)

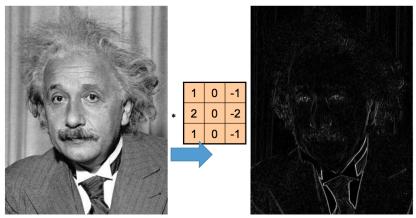
Idea: each neuron on the higher layer is detecting the same feature, but in different locations on the lower layer

The red connections all have the same weight.



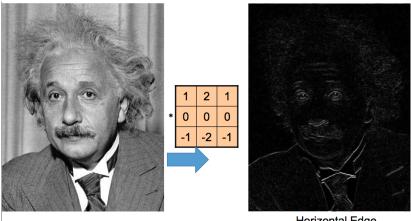
"Detecting" =the output (activation) is high if the feature is present

# Sobel Filter - Weights to Detect Vertical Edges



Vertical Edge (absolute value)

# Sobel Filter - Weights to Detect Horizontal Edges



Horizontal Edge (absolute value)<sup>8</sup>

#### Weights to Detect Blobs







#### Example:

Greyscale input image of size  $7\times7$  Convolution  $\mbox{kernel}$  of  $3\times3$  Questions:

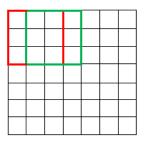
- How many units are in the higher layer?
- How many weights?

#### Example:

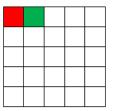
Greyscale input image of size  $7\times7$  Convolution  $\mbox{kernel}$  of  $3\times3$  Questions:

- How many units are in the higher layer?
- How many weights?

#### 7 x 7 Input Volume



#### 5 x 5 Output Volume



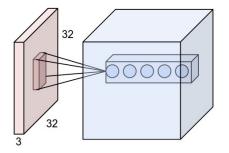
#### Computation:

http://deeplearning.stanford.edu/wiki/index.php/File:Convol

Colour input image of size 7  $\times$  7  $\times$  **3** Convolution kernel of 3  $\times$  3  $\times$  **3** Questions:

- How many units are in the higher layer?
- How many weights?

# **Detecting Multiple Features**



Input image of size  $7 \times 7 \times 3$  Convolution kernel of  $3 \times 3 \times 3 \times 5$ Questions:

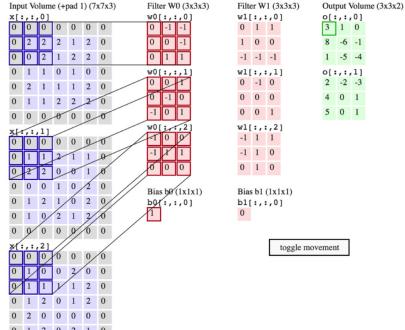
- How many units are in the higher layer?
- How many weights?

#### Example

Input image of size 32  $\times$  32  $\times$  3 Convolution kernel of 3  $\times$  3  $\times$  3  $\times$  10 Questions:

- How many units are in the higher layer?
- How many weights?

# Zero Padding



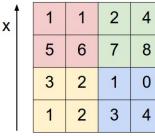
# Why Zero Padding

- Keep the hidden unit "width" & "height" consistent, or at least still divisible by 2.
- Keep the information around the border of the image.

# Max-Pooling

#### Single depth slice

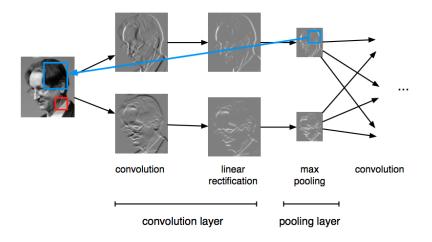
y



max pool with 2x2 filters and stride 2



# Max-Pooling Example



# Other Types of Pooling Operations:

- Average pooling (compute the average activation of a region)
- Max pooling generally works better
- More recently people are doing away with pooling operations (and using strided convolutions instead)

# Strided Convolution

#### 7 x 7 Input Volume

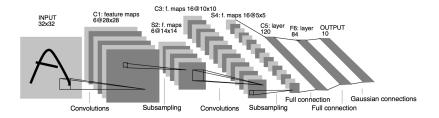
# 3 x 3 Output Volume



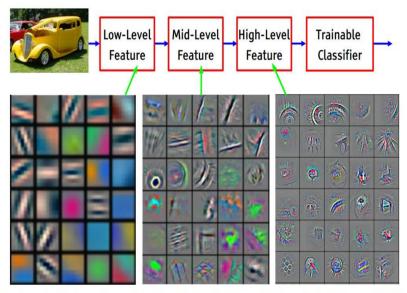
#### Animations

https://github.com/vdumoulin/conv\_arithmetic

#### LeNet Architecture



# What features do CNN's detect?



Feature visualization of convolutional net trained on ImageNet from [Zeiler & Fergus 2013]

Convolutional Networks in PyTorch

Pytorch!