



NTNU

Norwegian University of
Science and Technology

DEEP LEARNING CONVOLUTIONAL NEURAL NETWORKS

Tor Andre Myrvoll
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Introduction - Outline

Convolutional Neural Networks (CNNs)

- ▶ Motivation
- ▶ Basic CNNs
- ▶ Transfer learning

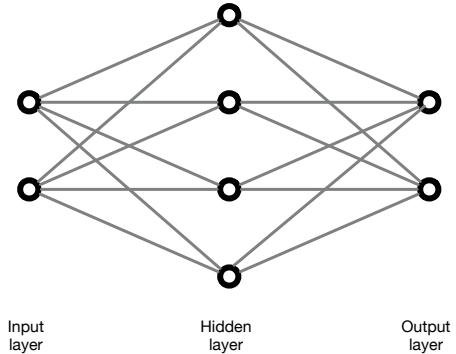
Introduction - Motivation



Introduction - Motivation

MLP Specification

- ▶ Binary classifier
- ▶ 512×512 images
- ▶ One hidden layer with as many nodes as there are inputs



Introduction - Motivation

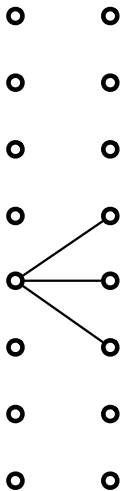
Reality catches up...

- ▶ We stack the image into a columnvector of dimension $512 \times 512 = 262144$. Large, but doable
- ▶ The number of hidden nodes will also be 262144 dimensions
- ▶ The number of connections from input to output will be $262144^2 = 68719476736$
- ▶ Which is roughly 268 GB of data is using float
- ▶ Game over?

Introduction - Solutions

Only use a subset of the connections

- ▶ Using $k \ll N$ inputs will decrease number of parameters
- ▶ Ex. $k = 3 \Rightarrow k \cdot N = 786432$, a reduction of factor 90000



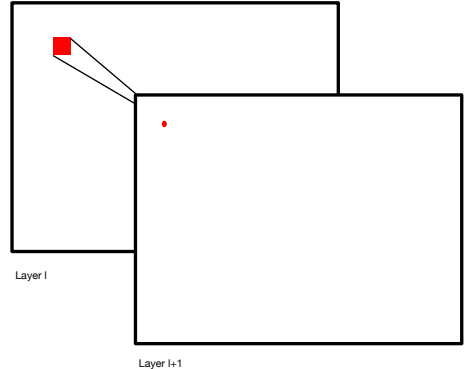
Introduction - The CNN

For images it makes sense to use local information

- ▶ The excitation in the $l + q$ layer is given as

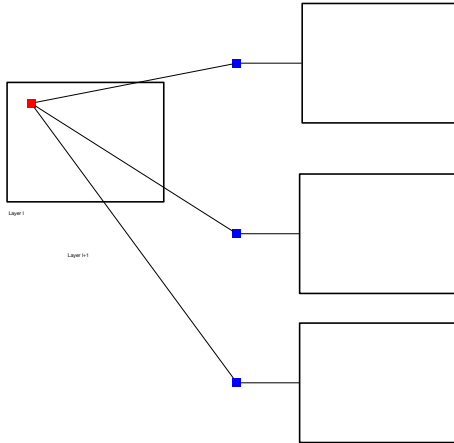
$$z_{i,j}^{l+1} = \sum_{m,n=-L/2}^{L/2} k_{m,n} v_{i-m,j-n}^l$$

- ▶ We see that this corresponds to a *convolution*, hence the naming scheme



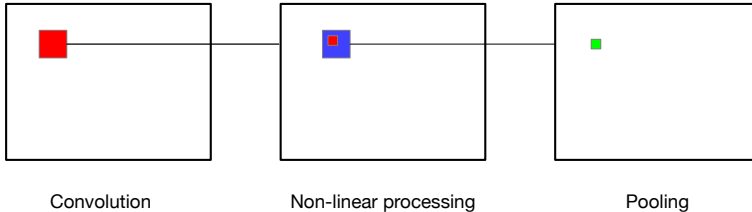
Introduction - The CNN

We can use multiple filters



Introduction - The CNN

In addition to convolution we do non-linear processing and *pooling*



Introduction - The CNN

The three processing steps that defines a CNN *layer* are:

- ▶ Convolution

$$z_{i,j}^{l+1} = \sum_{m,n=-L/2}^{L/2} k_{m,n} v_{i-m,j-n}^l$$

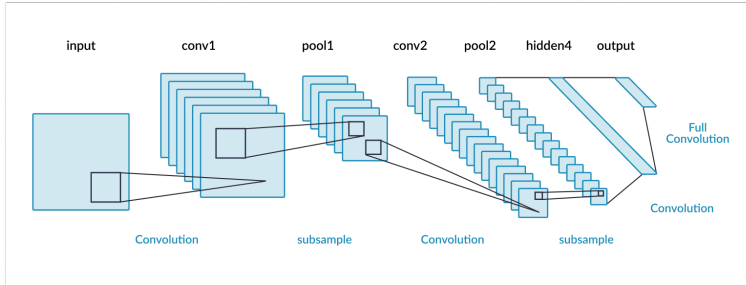
- ▶ Non-linear processing: ReLU, Sigmoid, etc.
- ▶ Pooling: Representing local areas by some statistic
 - ▶ Examples: Mean value, median, maximum value, power

$$z_{i,j}^l = \max\{v_{i-m,j-n}^l\}_{m,n=-L/2}^{L/2}$$

- ▶ Pooling leads to dimensionality reduction, and implicit translation and rotational invariance

Introduction - CNN example - LeNet

Early example of CNN – LeNet-5. Handwritten character recognition:



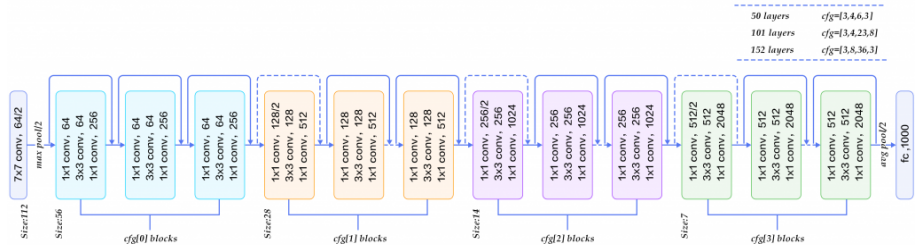
Introduction - CNN example - LeNet

Structure:

- ▶ Layer 1: Convolution. Six 5×5 kernels
- ▶ Layer 2: Pooling: $32 \times 32 \rightarrow 14 \times 14$
- ▶ Layer 3: Convolution. 16 5×5 kernels
- ▶ Layer 4: Pooling: $10 \times 10 \rightarrow 5 \times 5$
- ▶ Layer 5: Convolution: 120 5×5 kernels
- ▶ Layer 6: Fully connector. 84 features output.

Introduction - CNN example - ResNet

Massive 152 layer network for image classification



Introduction - Transfer learning

Training a large CNN is a massive undertaking. Use *transfer learning* to use a previously training CNN.

- ▶ Only use the CNN layers of a model
- ▶ This part of the model is often called a *feature extractor*
- ▶ Define and train your own fully connected layers after the CNN (don't update the CNN parameters)
- ▶ Available from for instance the Keras Applications API
- ▶ Available from Apple when using the COre ML API

Thank you for your attention

