

CONVOLUTIONAL NEURAL NETWORKS - I

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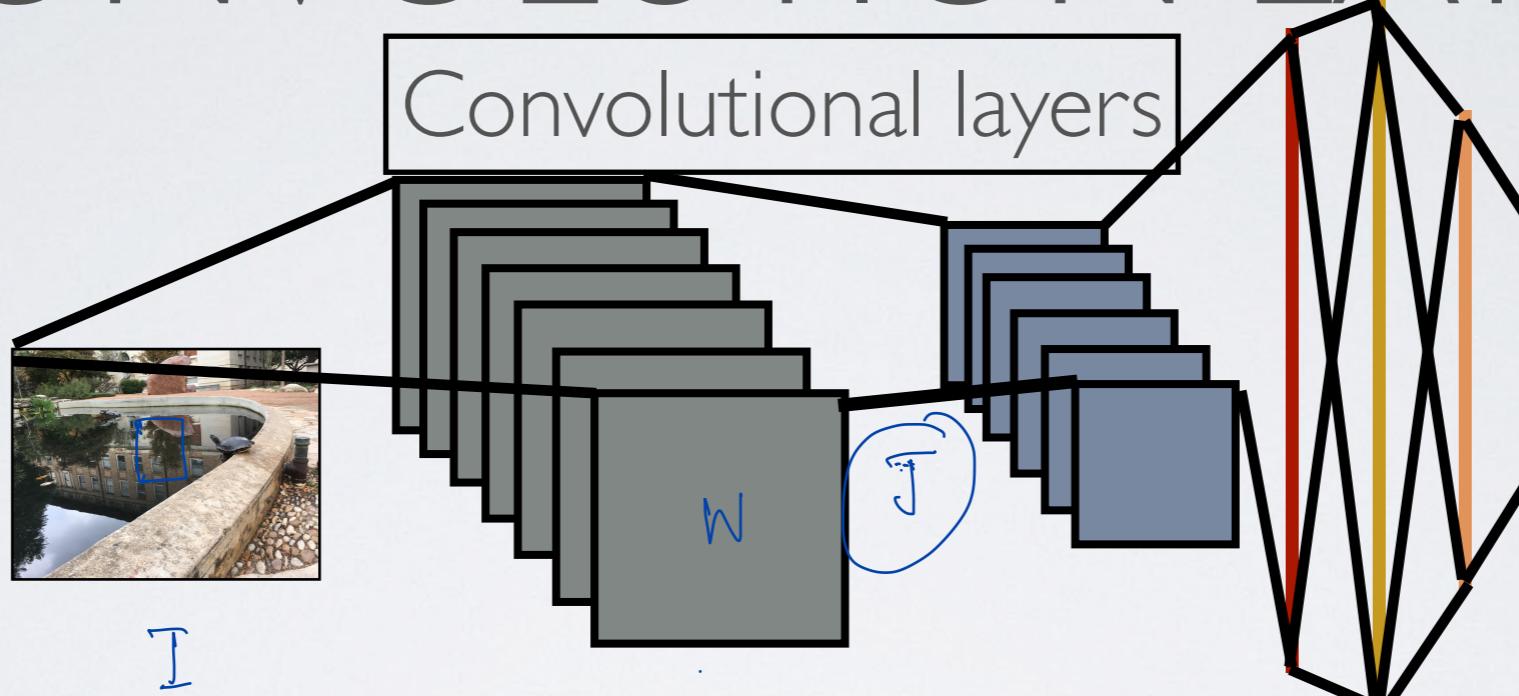
OUTLINE

- Motivation
- Building blocks
 - Convolutional layer/non-linearity layer
 - Pooling layer
 - Fully connected layer
- Training
 - Loss function
 - Gradient descent and learning rate
 - Data handling and stopping condition
- Generalisation

MOTIVATION

- Human Visual System (HVS)
- Dealing with 2D data 
- Parameter reduction 

BUILDING BLOCK: CONVOLUTION LAYER



- Linear component of the network
- Parameter sharing \approx
- Local connectivity \approx

$$J[p, q] = \sum_{m} \sum_{n} w[m, n] I[p-m, q-n]$$

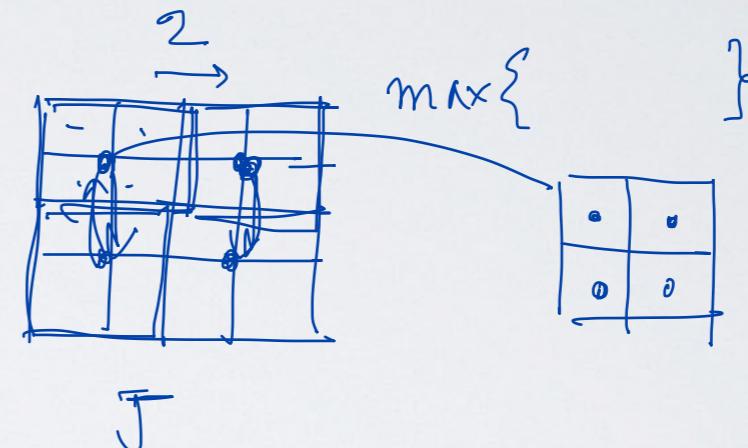
(NL) Conv (linear)

BUILDING BLOCK: NON-LINEARITY LAYER

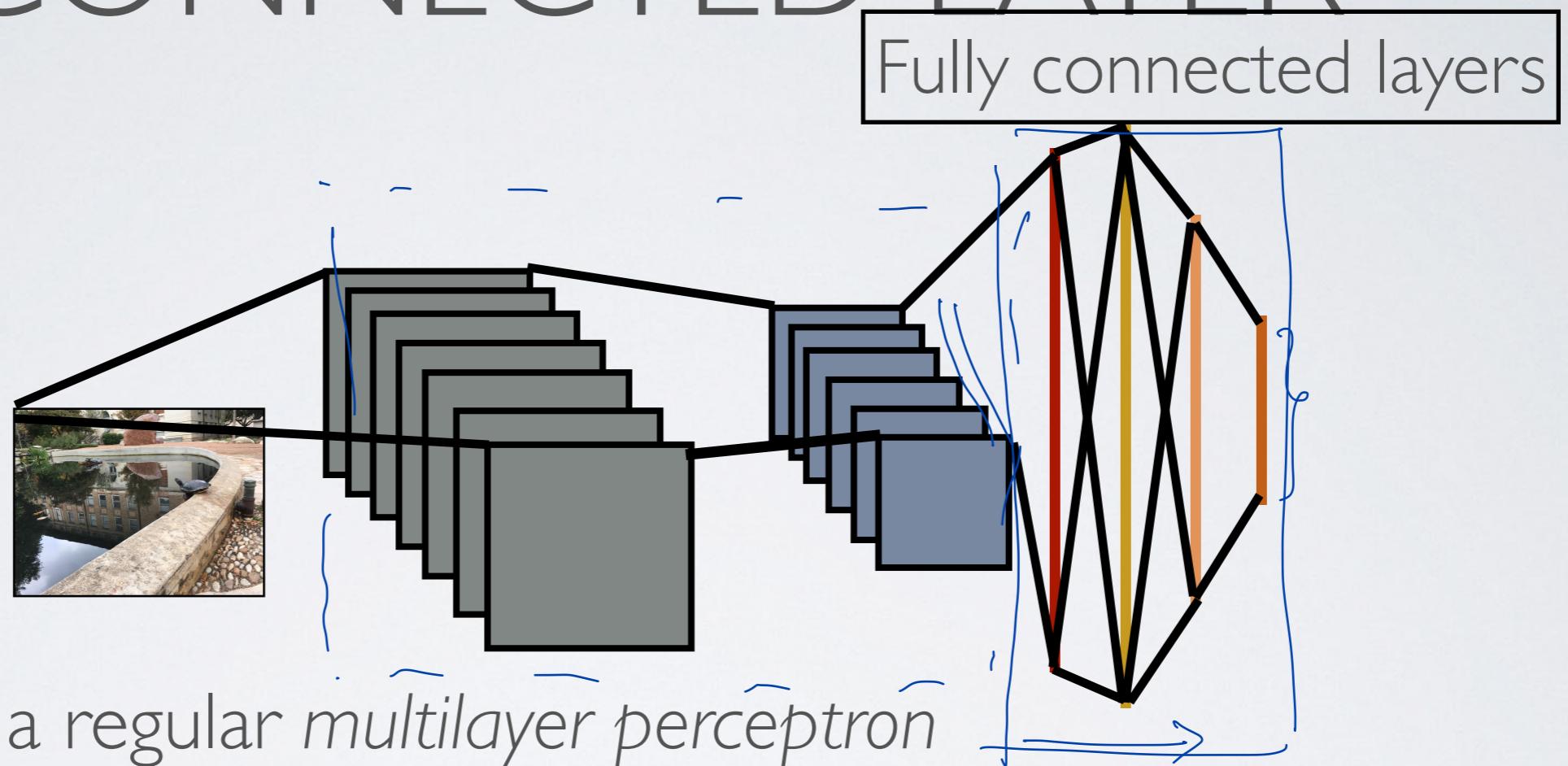
- Fundamental to NN
 - Allows for modelling complex *input/output relations*
 - Inspired by the *neuron*
 - Examples: sigmoid, \tanh , ReLU etc.
 - Also called Activation Layer

BUILDING BLOCK: POOLING LAYER

- Pooling of output
 - Dimensionality reduction
 - Control overfitting //
 - Invariant to small translation
 - Examples: max, l2 norm, average etc.



BUILDING BLOCK: FULLY CONNECTED LAYER



- Essentially, a regular *multilayer perceptron*
- Input is a vector formed by *flattening* pooling layer output
- Output is vector whose size equals *number of classes*

TRAINING: LOSS FUNCTION

- Standard loss functions such as softmax or cross entropy
- Loss functions are *non-convex* leading to *locally optimal* solutions

TRAINING: GRADIENT DESCENT BASED METHODS

- Several optimisation methods can be applied:
 - Gradient descent, Stochastic gradient descent
 - Momentum, Nesterov momentum $\mathcal{N}^{(r)}$
 - AdaGrad, RMSProp, Adam
- Stopping condition based on training/validation error
- Choice of *initialisation* important
- Backpropagation of forward loss used for gradient computation

TRAINING: PARAMETERS AND DATA

- Similar rules as in neural network training:
 - Weights typically *initialised randomly*
 - *Pre-trained weights* also used commonly
 - Feed data in *small batches* (*mini batches*)
 - *Epoch* is one forward and backward pass of all training data points

GENERALISATION

- Any model's performance is measured on *previously unseen inputs*
- This is also known as *generalisation*
- Typically, the dataset is divided into *training, validation and test set*
- Generalisation performance is reported on the *test set*

SUMMARY

- CNNs form the *fundamental building blocks* of modern machine learning models
- Allow for feature learning
- *Extremely successful* in solving several machine learning problems: image recognition, video analysis, natural language processing, drug discovery, visual system modelling etc.
- Reason for the *deep learning revolution*

REFERENCES

- <https://www.deeplearningbook.org>
- <https://cs231n.github.io/convolutional-networks/>