

Hands-on Deep Learning for Computer Vision and Biomedicine

Practical Course
Winter Semester 2018/2019

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Learning Goals

- Theory & Practice:
 - Basics and advanced techniques
- Deep learning **craftsmanship**
 - Understanding practical problems
 - Designing solutions
- Practical project experience with **real-world open problems**
 - The projects are geared towards producing scientific publications
- Presentation skills

Structure of Practical Course

- Three lectures in the beginning of the semester (Tuesday 2-4pm)
- Practical project
 - Own project for each group
 - 1 or 2 students per group
 - Most projects: Python, NumPy, deep learning frameworks
 - Access to computers and GPUs in Garching and remotely
 - Deep learning requires early and regular efforts
 - Regular communication with supervisors (important for progress of learning and project success)
- Final presentations
 - Groups can learn from each other and discuss
 - Presentation dates will be determined by voting (end of semester)

Next Steps

- **29 June - 4 July:** Apply for a place at <https://matching.in.tum.de/>
- There are many applicants
- Sending info about yourself is crucial to get matched and to get assigned a project with appropriate difficulty
- Email us info **until 6 July:**
 - Your programming skills
 - Some code you wrote in any context
 - Your interests, learning goals
 - Your courses, grades
- If you require project info in advance, contact us
- If you want to propose own projects ideas, they should be discussed with us until **6 July**
- Places in the course will be assigned on **12 July**

After **12 July**

- Projects will be announced, discussed and assigned as soon as possible
- We will consider your preferences, and also our knowledge about which of your preferred projects match your programming skills

Most Importantly

- Most importantly:
 - Read project descriptions very carefully, select projects wisely
 - Follow all announced rules

Other Options

- If you don't get a place in the practical course:
 - Email us, enter the waiting list
 - Apply in subsequent semesters
- Whether you get a place or not, also consider applying for:
 - Bachelor Thesis
 - Master Thesis
 - Interdisciplinary Project
 - Guided Research
 - etc.

Prerequisites

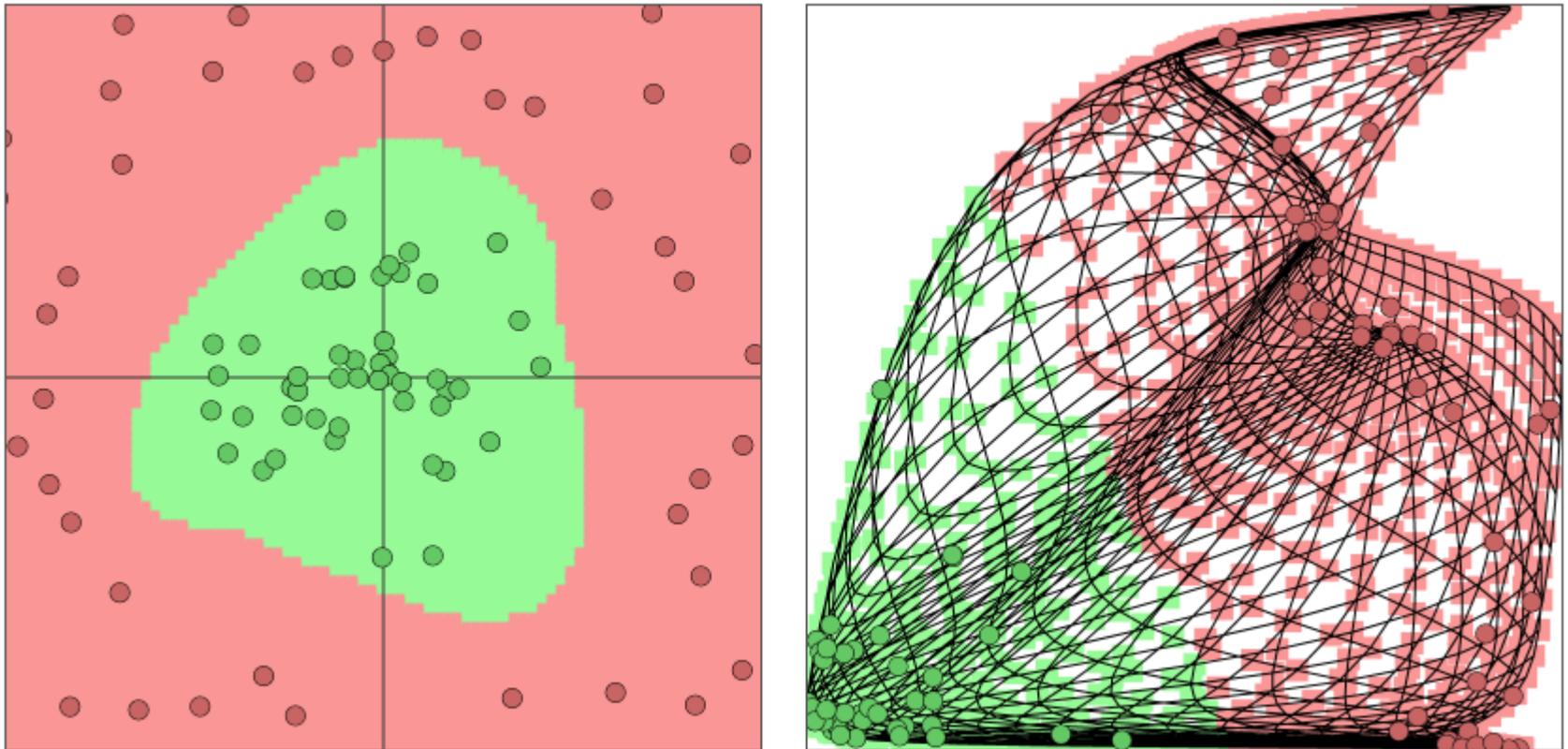
- Good programming skills
 - Python
 - Array programming in NumPy (or Matlab or similar)
- Curiosity
- Time for regular hard work
- Soft skills

- In many of the projects, prior knowledge (about deep learning, computer vision, biomedicine) is **not** required
 - You will learn from your supervisor
- But good programming skills are important

Literature

- Christopher M. Bishop: “Pattern Recognition and Machine Learning”, Springer, 2006.
- <http://www.deeplearningbook.org/>
- <http://neuralnetworksanddeeplearning.com/>
- <http://www.mlyearning.org/>
- NumPy: Advanced Array Indexing
<https://docs.scipy.org/doc/numpy/reference/arrays.indexing.html>

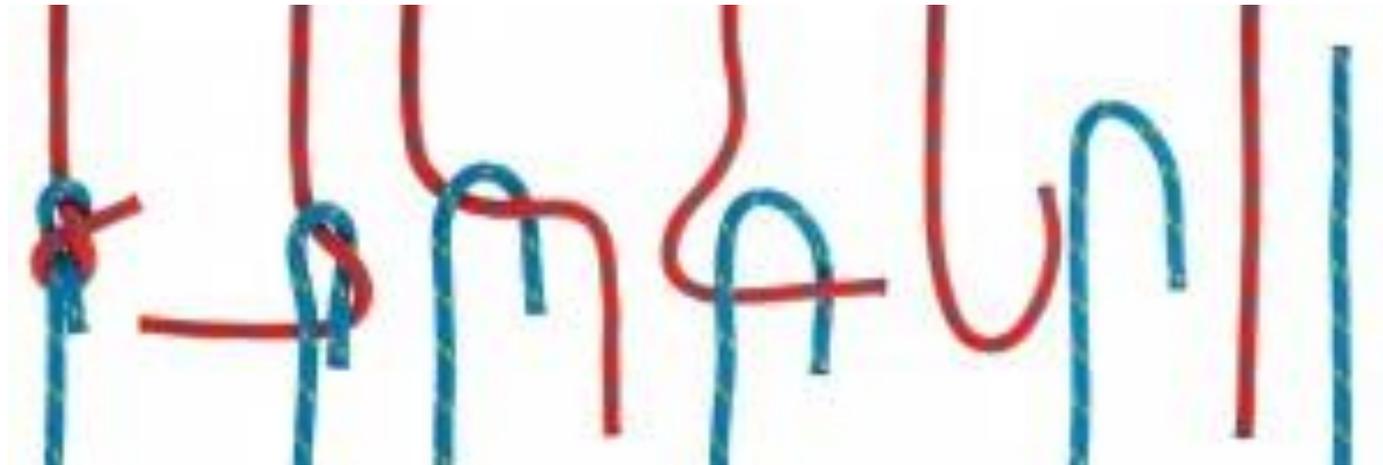
Nonlinear Coordinate Transformation



<http://cs.stanford.edu/people/karpathy/convnetjs/>

Dimensionality may change! (Here: 2D to 2D)

Deep Neural Network: Sequence of Many Simple Nonlinear Coordinate Transformations that “disentangle” the data (by transforming the entire coordinate system)



Data is sparse (almost lower-dimensional)

Linear separation of red and blue classes

Fully-Connected Layer a.k.a. Dense Layer

$x^{(0)}$ is input feature vector for neural network (one sample).

$x^{(L)}$ is output vector of neural network with L layers.

Layer number l has:

- Inputs (usually $x^{(l-1)}$, i.e. outputs of layer number $l - 1$)
- Weight matrix $W^{(l)}$, bias vector $b^{(l)}$ - both trained (e.g. with stochastic gradient descent) such that network output $x^{(L)}$ for the training samples minimizes some objective (loss)
- Nonlinearity s_l (fixed in advance, for example $\text{ReLU}(z) := \max\{0, z\}$)
- Output $x^{(l)}$ of layer l

Transformation from $x^{(l-1)}$ to $x^{(l)}$ performed by layer l :

$$x^{(l)} = s_l \left(W^{(l)} x^{(l-1)} + b^{(l)} \right)$$

One Layer: Graphical Representation

$$W^{(l)} = \begin{pmatrix} 0 & 0.1 & -1 \\ -0.2 & 0 & 1 \end{pmatrix}$$

$$x^{(l-1)} = \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix}$$

$$b^{(l)} = \begin{pmatrix} 0 \\ 1.2 \end{pmatrix}$$

$$W^{(l)}x^{(l-1)} + b^{(l)} =$$

$$= \begin{pmatrix} 0 \cdot 1 + 0.1 \cdot 2 - 1 \cdot 3 + 0 \\ -0.2 \cdot 1 + 0 \cdot 2 + 1 \cdot 3 + 1.2 \end{pmatrix}$$

$$= \begin{pmatrix} -2.8 \\ 4 \end{pmatrix}$$

