EE-559



Deep learning

Cursus	Sem.	Туре
Civil & Environmental Engineering		Opt. Language
Computational science and Engineering	MA2, MA4	Opt.
Computer science	MA2, MA4	Opt.
Cybersecurity	MA2, MA4	Opt.
Data Science	MA2, MA4	4 Opt. Workload Weeks
Data science minor	E	Opt.
Digital Humanities	MA2, MA4	Opt.
lectrical Engineering		Opt.
lectrical and Electronical Engineering	MA2, MA4	Opt.
inancial engineering	MA2, MA4	Opt.
earning Sciences		Obl.
ife Sciences Engineering	MA2, MA4	Opt.
linor in Quantum Science and Engineering	E	Opt.
leuro-X minor	E	Opt.
Neuro-X	MA2	Opt.
Quantum Science and Engineering	MA2	Opt.
Robotics, Control and Intelligent Systems		Opt.
Robotics	MA2, MA4	Opt.
SC master EPFL	MA2, MA4	Opt.
Statistics	MA2	Opt.

Remark

pas donné en 2022-23

Summary

The objective of this course is to provide a complete introduction to deep machine learning. How to design a neural network, how to train it, and what are the modern techniques that specifically handle very large networks.

Content

The course aims at providing an overview of existing processings and methods, at teaching how to design and train a deep neural network for a given task, and at providing the theoretical basis to go beyond the topics directly seen in the course.

It will touch on the following topics:

- What is deep learning, introduction to tensors.
- Basic machine-learning, empirical risk minimization, simple embeddings.
- Linear separability, multi-layer perceptrons, back-propagation.
- Generalized networks, autograd, batch processing, convolutional networks.
- Initialization, optimization, and regularization. Drop-out, batchnorm, resnets.
- Deep models for Computer Vision.
- Analysis of deep models.

- Auto-encoders, embeddings, and generative models.
- Recurrent and attention models, Natural Language Processing.

Concepts will be illustrated with examples in the PyTorch framework (http://pytorch.org).

Keywords

machine learning, neural networks, deep learning, computer vision, python, pytorch

Learning Prerequisites

Required courses

- Linear algebra (vector, matrix operations, Euclidean spaces).
- Differential calculus (Jacobian, Hessian, chain rule).
- Python programming.

• Basics in probabilities and statistics (discrete and continuous distributions, normal density, law of large numbers, conditional probabilities, Bayes, PCA)

Recommended courses

- Basics in optimization (notion of minima, gradient descent).
- Basics in algorithmic (computational costs).
- Basics in signal processing (Fourier transform, wavelets).

Teaching methods

Ex-cathedra with exercise sessions and mini-projects. Possibly invited speakers.

Assessment methods

Mini-projects by groups of students, and one final written exam.

Resources

Notes/Handbook Not mandatory: http://www.deeplearningbook.org/

Websites

• https://fleuret.org/ee559/

Moodle Link

• https://go.epfl.ch/EE-559