

# EE-559 Deep learning

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Cursus	Sem.	Type
Electrical and Electronical Engineering	MA2, MA4	Opt.

Language of **English** teaching Credits Session Summer Semester Spring Exam Written Workload 120h Weeks 14 Hours 4 weekly 2 weekly Courses Exercises 2 weekly Number of positions

### **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 teaching the required skills to use deep learning methods on applied problems. It will show how to design and train a deep neural network for a given task, and the sufficient theoretical

It will show how to design and train a deep neural network for a given task, and the sufficient theoretical basis to go beyond the topics directly seen in the course.

The planned content of the course:

- What is deep learning, introduction to tensors.
- Basic machine-learning, empirical risk minimization, simple embeddings.
- Linear separability, multi-layer perceptrons, back-prop.
- Generalized networks, autograd, batch processing, convolutional networks.
- Initialization, optimization, and regularization. Drop-out, activation normalization, skip connections.
- Deep models for Computer Vision.
- Analysis of deep models.
- Auto-encoders, embeddings, and generative models.
- Recurrent models and Natural Language Processing.
- pytorch tensors, deep learning modules, and internals.

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).

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- 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).

#### **Learning Outcomes**

By the end of the course, the student must be able to:

- Design a deep neural network
- · Design a training procedure
- Design a sound evaluation protocol

#### Transversal skills

- Use a work methodology appropriate to the task.
- · Collect data.
- Demonstrate a capacity for creativity.
- Continue to work through difficulties or initial failure to find optimal solutions.

## **Teaching methods**

Ex-cathedra with exercise sessions and mini-projects. Invited speakers from the industry will present how deep learning is used in practice for their applications.

### **Assessment methods**

Two mini-projects by groups of three students, and one final written exam.

### Supervision

Office hours No
Assistants Yes
Forum Yes

### Resources

### Ressources en bibliothèque

• Deep Learning / Goodfellow

# Notes/Handbook

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

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