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Cursus	Sem.	Type
Computer science	MA2	Opt.
Cybersecurity	MA2	Obl.
Data Science	MA2, MA4	Opt.
SC master EPFL	MA2, MA4	Opt.

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

#### Summary

Machine learning and data analysis are becoming increasingly central in many sciences and applications. This course concentrates on the theoretical underpinnings of machine learning.

#### Content

- Basics: statistical learning framework, Probably Approximately Correct (PAC) learning, learning with a finite number of classes, Vapnik-Chervonenkis (VC) dimension, non-uniform learnability, complexity of learing.
- Neural Nets: representation power of neural nets, learning and stability, PAC Bayes bounds.
- · Graphical model learning.
- Non-negative matrix factorization, Tensor decompositions and factorization.
- · Learning mixture models.

## **Learning Prerequisites**

### **Recommended courses**

- Analysis I, II, III
- Linear Algebra
- Machine learning
- Probability
- Algorithms (CS-250)

#### **Learning Outcomes**

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

- Explain the framework of PAC learning
- Explain the importance basic concepts such as VC dimension and non-uniform learnability
- Describe basic facts about representation of functions by neural networks
- Describe recent results on specific topics e.g., graphical model learning, matrix and tensor factorization, learning mixture models

# **Teaching methods**

Learning theory Page 1 / 2



- Lectures
- Exercises

# **Expected student activities**

- Attend lectures
- Attend exercises sessions and do the homework

# **Assessment methods**

Final exam and graded homeworks

# Supervision

Office hours Yes
Assistants Yes
Forum Yes

Others Course website

Learning theory Page 2 / 2