

# EE-613 Machine Learning for Engineers

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Cursus	Sem.	Type
Civil & Environmental Engineering		Obl.
Electrical Engineering		Opt.
Robotics, Control and Intelligent Systems		Opt.

Language of teaching	English
Credits	4
Session	
Exam	Multiple
Workload	120h
Hours	56
Courses	28
TP	28
Number of positions	40

# Frequency

Every 2 years

#### Remark

Next time: Fall 2023

# **Summary**

The objective of this course is to give an overview of machine learning techniques used for real-world applications, and to teach how to implement and use them in practice. Laboratories will be done in python using jupyter notebooks.

#### Content

### Fundamentals

- Recalls in probability and information theory
- Notion of learning, cross validation and performance evaluation
- Optimization (gradient, Newton method, stochastic gradient, etc.)

#### Regression

- Least squares (Tikhonov regularization)
- Weighted least squares
- Iteratively reweighted least squares (IRLS)
- Tensor regression
- Gaussian mixture regression (GMR)
- Gaussian process regression (GPR)

### Generative models

- · Bayesian networks,
- Directed / non-directed models, conditional independence, Naive Bayes
- k-Means, Gaussian mixture model (GMM), Expectation-Maximization (EM)
- PCA and probabilistic PCA
- Hidden Markov model (HMM)

### Discriminative models



- · Logistic regression
- · Decision trees, random forest
- Support vector machine (SVM) and Kernelization (PCA, etc.)

#### Deep learning

- Perceptron, Multi-Layer-Perceptron (MLP)
- Convolutional neural network (CNNs)

# Meta-algorithms

- Ensemble methods
- Bagging
- Boosting

# **Learning Prerequisites**

#### Required courses

- Undergraduate knowledge of probabilities, linear algebra, and statistics
- Python programming

### **Learning Outcomes**

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

- Understand the core principles of machine learning and of the different concepts and algorithms behind the different learning methodologies.
- Select appropriately in practice standard learning-based inference techniques for regression, classification and density modeling, including understanding the impact of different parameters

#### **Assessment methods**

Multiple.

#### Resources

**Bibliography**Pattern Recognition and Machine Learning, C. Bishop,
Springer, 2008

### Ressources en bibliothèque

• Pattern Recognition and Machine Learning / Bishop