

## Course guide

### 230706 - DLAI - Deep Learning for Artificial Intelligence

Last modified: 25/05/2023

**Unit in charge:** Barcelona School of Telecommunications Engineering  
**Teaching unit:** 739 - TSC - Department of Signal Theory and Communications.

**Degree:** MASTER'S DEGREE IN TELECOMMUNICATIONS ENGINEERING (Syllabus 2013). (Optional subject).  
MASTER'S DEGREE IN ADVANCED TELECOMMUNICATION TECHNOLOGIES (Syllabus 2019). (Optional subject).

**Academic year:** 2023    **ECTS Credits:** 5.0    **Languages:** English

#### LECTURER

**Coordinating lecturer:** Consultar aquí / See here:  
<https://telecos.upc.edu/ca/estudis/curs-actual/professorat-responsables-coordinadors/responsables-assignatura>

**Others:** Consultar aquí / See here:  
<https://telecos.upc.edu/ca/estudis/curs-actual/professorat-responsables-coordinadors/professorat-assignat-idioma>

#### PRIOR SKILLS

A previous knowledge on basic machine learning is advisable. In terms of programming, it is recommended that students are familiar with Python programming language beforehand.

#### DEGREE COMPETENCES TO WHICH THE SUBJECT CONTRIBUTES

**Specific:**  
CE1. Ability to apply information theory methods, adaptive modulation and channel coding, as well as advanced techniques of digital signal processing to communication and audiovisual systems.

#### TEACHING METHODOLOGY

Lectures, in class labs and assignments.

#### LEARNING OBJECTIVES OF THE SUBJECT

At the end of this course students will be able to design, implement, train and evaluate a machine learning system based on deep neural networks.

#### STUDY LOAD

Type	Hours	Percentage
Hours small group	13,0	10.40
Self study	86,0	68.80
Hours large group	26,0	20.80

**Total learning time:** 125 h



## CONTENTS

### 1. DEEP NEURAL NETWORKS

#### Description:

- 1.1 The Perceptron. Regression vs classification. The Softmax classifier.
- 1.2 Multi-layer perceptron (MLP).
- 1.3 Basic layers: Fully connected. Convolutions/Deconvolutions, Non-linearities (ReLU, tanh, sigmoid). Downsampling/Upsampling.
- 1.4 Interpretability: t-SNE, visualizations, highest activations.

**Full-or-part-time:** 18h

Theory classes: 3h 57m

Self study : 14h 03m

### 2. TRAINING

#### Description:

- 2.1 Backpropagation
- 2.2 Optimizers
- 2.3 Loss functions
- 2.4 Methodology
- 2.5 Efficient computation

**Full-or-part-time:** 35h 59m

Theory classes: 7h 53m

Self study : 28h 06m

### 3. MEMORY NETWORKS

#### Description:

- 3.1 Recurrent Neural Networks
- 3.2 Gated models: LSTM, GRU, ...
- 3.3 Advanced models: QRNN, pLSTM, ...

**Full-or-part-time:** 18h

Theory classes: 3h 57m

Self study : 14h 03m

### 4. BEYOND SUPERVISED LEARNING

#### Description:

- 4.1 Unsupervised and semi-supervised learning.
- 4.2 Adversarial training and generative models
- 4.3 Incremental learning
- 4.4 Active learning
- 4.5 Reinforcement learning
- 4.6 Meta-learning

**Full-or-part-time:** 18h

Theory classes: 3h 57m

Self study : 14h 03m



## 5. COMPUTATION

### Description:

- 5.1 Software stack
- 5.2 Computational requirements
- 5.3 Scalability

**Full-or-part-time:** 18h

Theory classes: 3h 57m

Self study : 14h 03m

## ACTIVITIES

### Lectures

### Description:

1. DEEP NEURAL NETWORKS
2. TRAINING
3. MEMORY NETWORKS
4. BEYOND SUPERVISED LEARNING
5. COMPUTATION

**Full-or-part-time:** 108h

Theory classes: 23h 40m

Self study: 84h 20m

### Labs in class

### Description:

1. Classification vs Regression
2. Convolutional neural networks for image classification.
3. Data pipelines between CPUs and GPUs.
4. Interpretability of a convolutional neural network.
5. Generative adversarial networks.

### Material:

Deep learning frameworks used during the labs: Caffe, Tensorflow and Keras.

**Full-or-part-time:** 10h

Laboratory classes: 5h

Self study: 5h



### Project

**Description:**

Hands on project where students must design, train and test their own deep learning model.

**Material:**

GPUs on a cloud service.

**Delivery:**

Oral presentation

Poster

**Full-or-part-time:** 40h

Theory classes: 1h

Laboratory classes: 8h

Self study: 31h

### Grading

**Description:**

Written exams in class.

**Full-or-part-time:** 4h

Theory classes: 4h

## GRADING SYSTEM

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Labs: 15%

Midterm: 15%

Project: 40%

Final exam: 30%

## BIBLIOGRAPHY

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**Basic:**

- Goodfellow, Ian; Bengio, Yoshua; Courville, Aaron. Deep learning [on line]. Boston: MIT Press, 2016 [Consultation: 16/06/2017]. Available on: <http://www.deeplearningbook.org/>. ISBN 978-0262035613.

## RESOURCES

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**Audiovisual material:**

- <https://telecombcn-dl.github.io/2017-dlsl/>. Resource

**Hyperlink:**

- <https://telecombcn-dl.github.io/2017-dlcv/>. Deep Learning for Computer Vision Summer School at UPC ETSETB TelecomBCN 2017

- <https://telecombcn-dl.github.io/2017-dlai/>. Web page of the course