

Deep Learning

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Source: course of François Fleuret at EPFL

<https://fleuret.org/dlc/> (cc-by-nc-sa)

Plan du cours

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Séance 1

Introduction

- 1.1. – From neural networks to deep learning. (18 slides) ([slides](#))
- 1.2. – Current applications and success. (25 slides) ([slides](#))
- 1.3. – What is really happening? (10 slides) ([slides](#))
- 1.4. – Tensor basics and linear regression. (13 slides) ([slides](#))
- 1.5. – High dimension tensors. (20 slides) ([slides](#))

From perceptron to deep convolutional neural networks

- 3.1. – The perceptron. (16 slides) ([slides](#))
- 3.3. – Linear separability and feature design. (10 slides) ([slides](#))
- 3.4. – Multi-Layer Perceptrons. (10 slides) ([slides](#))
- 4.4. – Convolutions. (23 slides) ([slides](#))
- 4.5. – Pooling. (7 slides) ([slides](#))
- 8.2. – Networks for image classification. (34 slides) ([slides](#))

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Séance 2: Training deep neural networks

Back-propagation

- 3.5. – Gradient descent. (13 slides) ([slides](#))
- 3.6. – Back-propagation. (11 slides) ([slides](#))
- 4.1. – DAG networks. (11 slides) ([slides](#))

Initialization, optimization and regularization

- 5.1 – Cross-entropy loss. (9 slides) ([slides](#))
- 5.2 – Stochastic gradient descent. (17 slides) ([slides](#))
- 5.5 – Parameter initialization. (22 slides) ([slides](#))
- 5.6 – Architecture choice and training protocol. (9 slides) ([slides](#))
- 6.1 – Rectifiers. (7 slides) ([slides](#))
- 6.3 – Dropout. (12 slides) ([slides](#))
- 6.4 – Batch normalization. (15 slides) ([slides](#))

TD: [Practical 3](#) - implement a MLP with one hidden layer

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Séance 3: Pytorch practicals

Pytorch NN implementation

4.2 – Autograd. (9 slides) ([slides](#))

4.3 – PyTorch modules and batch processing. (5 slides) ([slides](#))

4.6 – Writing a PyTorch module. (10 slides) ([slides](#))

5.3 – PyTorch optimizers. (8 slides) ([slides](#))

5.7 – Writing an autograd function. (7 slides) ([slides](#))

6.6 – Using GPUs. (19 slides) ([slides](#))

TP: [Practical 4](#) - Implement a convolutional network