

DOCTORAL SCHOOL OF INFORMATICS
COMPLEX EXAM SUBJECT

Neural Computations (recommended subject)

1. Neural network structures
 - 1.1. Models: self-organized and supervised learning
 - 1.2. Principles of Hebbian learning, roles of graceful degradation, forgetting, normalization
 - 1.3. The basis of the variational principle (parametric approximation)
 2. Special networks
 - 2.1. Perceptron and its types (sigmoid, ReLU, etc.)
 - 2.2. Autoencoder, recurrent networks
 - 2.3. Hopfield network
 - 2.4. Generative networks
 3. Learning and training
 - 3.1. Stacked autoencoder, denoising autoencoder
 - 3.2. Backpropagation as a modified Hebbian learning
 - 3.3. Against overtraining: dropout, validation early stopping, and testing
 4. Other networks
 - 4.1. Support vector machines
 - 4.2. Sparse representations
 - 4.3. Convolutional neural networks, the role of convolution and pooling, connection to sparsity and to Self-Organizing Maps
 5. Optimization methods
 - 5.1. Types of cost functions
 - 5.2. Detection, Classification
 - 5.3. Function approximation
 - 5.4. Clustering and low dimensional embedding
-

Literature:

Goodfellow, I, Bengio, Y, and Courville, A. (2016) Deep Learning, MIT Press, Cambridge, MA. ISBN-10: 0262035618

Murphy, Kevin P. (2012) Machine Learning: A Probabilistic Perspective. MIT Press, Cambridge, MA. ISBN 978-0-262-01802-9

Haykin, S. (1999) Neural Networks: A Comprehensive Foundation, Prentice Hall, ISBN 0-13273350-1

Online materials:

<http://deeplearningbook.org>

<https://github.com/terryum/awesome-deep-learning-papers>

<https://www.coursera.org/learn/neural-networks>