

## Submodule MIN-336-01 Deep Learning

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| <b>Subheading</b>                              | (MIN-DL)                                       |
| <b>Person in Charge</b>                        | Pigors, Adrian, Prof. Dr.                      |
| <b>Language of Instruction</b>                 | by agreement                                   |
| <b>Curriculum Allocation</b>                   | MIN  |
| <b>Course Type, Contact Hours per Week</b>     | Lecture with exercise, 4 SWS                   |
| <b>ECTS Credits</b>                            | 6  |
| <b>Contact Hours / Independent Study Hours</b> | 68 h / 112 h                                   |
| <b>Suggestions for Independent Study</b>       | See bibliography                               |
| <b>Recommended Prerequisites</b>               | MIN-335  |
| <b>Examination</b>                             | Written or oral examination, experimental work |
| <b>Group Size</b>                              | 30   |

### Learning Outcomes

Algorithmic and mathematical skills: Students are able to explain the various models and algorithms of deep learning (DL) and describe their mathematical foundations.

Analysis, design, and realization skills: Students can select and combine suitable DL methods for a given problem. They can build and train neural networks and they can evaluate the quality of DL solutions.

Technological skills: Students are familiar with current DL frameworks and libraries.

### Content

Selected topics from the following areas:

- Basics of deep learning
- Convolutional neural networks (CNNs) for image data
- Recurrent neural networks (RNNs) and LSTM networks for sequential data
- Implementation of various network architectures in current software environments (e.g. Python, Keras, TensorFlow)
- Training of neural networks (hyperparameter optimization, regularization, etc.), transfer learning
- Deep learning for computer vision (CV): classic and modern CNN architectures for CV tasks such as image classification and object detection, autoencoders
- Deep learning for natural language processing (NLP): statistical language models, word embeddings, RNN and CNN architectures for NLP tasks such as text classification and generation, attention mechanism
- Current developments and applications of deep learning

### Requirements for Contact Hours

Active participation, solving exercise problems

### Requirements for Independent Study Hours

Preparation and review of the lectures

### Bibliography

Goodfellow, Bengio, Courville: Deep Learning. MIT Press, 2016

Chollet: Deep Learning with Python. Manning, 2020

Géron: Hands-On Machine Learning with Scikit-Learn, Keras, and TensorFlow. O'Reilly, 2019

Further current literature on the lecture contents