Introduction to Data Science for the Humanities

PhD in Philosophy and Human Sciences

Data analytics and machine learning

Lecture 1: Introduction to Data Science

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May 25, room Martinetti, 10:30 - 12:30
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A brief history of Artificial Intelligence and data science. The Data revolution. Models of machine learning. Deep learning. Ethical and social issues.

Lecture 2: The algorithmic tools of machine learning

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May 30, room Martinetti, 10:30 - 12:30
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Statistical learning. From data to their mathematical representation. Probabilistic models vs the vector space. Examples of image and textual encoding. Introduction to linear transformation and neural networks.

Lecture 3: Unsupervised learning

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June 6, room Martinetti, 10:30 - 12:30
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Principles of unsupervised learning. KMeans, an example of a clustering algorithm. Case study on image clustering.

Lecture 4: Supervised learning

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June 9, room Martinetti, 10:30 - 12:30
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Principles of supervised learning. Working with textual data. Intuition of language modeling and Recurrent Neural Networks. A case study on author prediction and text generation.

Lecture 5: Reinforcement learning and evolving neural networks

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June 13, room Martinetti, 10:30 - 12:30
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Intuition of learning by reinforcement. Differences between reinforcement learning and evolution. A case study on simulating natural and artificial selection.

Lecture 6: Deep Learning and introduction to Language Models

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June 16, room Martinetti, 10:30 - 12:30
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Introduction to Language Models. Statistical vs Neural language models. Main tasks that can be addressed using LMs.

Social and Ethical Issues in NLP

Lecture 7: Introduction to Large Language Models (LLMs)

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June 20, room Martinetti, 10:30 - 12:30
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Neural Language Models. Sequence to Sequence learning, Recurrent Neural Networks (RNN), Encoder-Decoder architectures, Attention and Transformers.

Lecture 8: Black Box models and Explainable Al

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June 23, room Martinetti, 10:30 - 12:30
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Introduction to the problem of explainable AI. Explainability and Causality. The role of Attention in the explanation of LLMs.

Lecture 9: Introduction to Social and Ethical Bias

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June 27, room Martinetti, 10:30 - 12:30
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Algorithmic bias. Hate-speech, toxicity. Privacy violation and profiling. Misinformation, fake-news, information and opinion manipulation. Technological divide.

Lecture 10: Automatic Detection of Implicit Bias and Stereotypes

June 30, room Martinetti, 10:30 - 12:30

Overview of the literature on the main approaches to automatic detection of bias and stereotypes.

References

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