

Master of Science (M.Sc.)

„Mannheim Master in Data Science“

University of Mannheim

– Module catalog –

Appendix

for students starting in or after autumn 2024

Academic Year

HWS 24/25

Die folgenden Veranstaltungen wurden nach Veröffentlichung des Modulkatalogs dem Kursprogramm hinzugefügt.

C. Data Management

| Module no. | Name of Module | Offered | Language | ECTS | Page |
|------------|------------------|---------|----------|------|------|
| IE 650 | Knowledge Graphs | HWS | E | 6 | 3 |

D. Data Analytics Methods

| Module no. | Name of Module | Offered | Language | ECTS | Page |
|------------|---|---------|----------|------|------|
| IE 695 | Reinforcement Learning | HWS | E | 6 | 6 |
| DS 204 | Mathematical Foundations of Machine Learning for Social data Scientists | FSS | E | 6 | |

G. Projects and Seminars

| Module no. | Name of Module | Offered | Language | ECTS | Page |
|------------|-----------------------------------|--------------|----------|------|------|
| MAS 515 | Seminar Mathematische Optimierung | unregelmäßig | D | 4 | 10 |

Detailed Descriptions

C. Data Management

| IE 650 | Knowledge Graphs |
|----------------|---|
| Form of module | Lecture |
| Type of module | Specialization course |
| Level | Master |
| ECTS | 6 |
| Workload | Hours per semester present at university: 56 h (4 SWS) |
| | Self-study: 124 h per semester <ul style="list-style-type: none">• 82 h: pre and post lecture studying and revision• 42 h: examination preparation |
| Prerequisites | Java or Python programming skills |
| Aim of module | <ul style="list-style-type: none">• The Role of knowledge graphs in the AI landscape• Semantic Web and its representation languages• Labeled property graphs• Query languages for knowledge graphs• Knowledge modeling and ontologies• Logical reasoning with knowledge graphs |

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| | <ul style="list-style-type: none"> Machine learning with knowledge graphs and knowledge graph embeddings |
| Learning outcomes and qualification goals | <p>Expertise:</p> <p>The participants of this course learn about principles and applications of knowledge graphs. They become familiar with their technical foundations such as representation and query languages, or logical inference. After taking this course, the students will be aware of the problems and benefits of knowledge graph technologies in the context of tasks such as knowledge management, information search and data integration, and they will be capable of judging the applicability of these technologies for addressing practical challenges.</p> <p>(MK1, MK2)</p> |
| | <p>Methodological competence:</p> <p>The participants learn how to design and implement AI systems based on knowledge graphs. They are able to use standardized modeling languages for building knowledge representations, and to query these models by means of languages such as SPARQL.</p> <p>(MF3)</p> |
| | <p>Personal competence:</p> <p>By jointly building a knowledge graph-based application, the students learn how to effectively work in teams. They improve upon their presentation skills by showing the outcomes of their projects to the other participants of the course.</p> <p>(MKO1, MKO3)</p> |
| Media | Lecture slides and exercise sheets will be available online |
| Literature | <ul style="list-style-type: none"> Pascal Hitzler, Markus Krötzsch and Sebastian Rudolph, Foundations of Semantic Web Technologies, Chapman & Hall/CRC, 2009 |

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| | <ul style="list-style-type: none"> • Allemang and Hendler (2008): Semantic Web for the Working Ontologist. Verlag Morgan Kaufmann. • Antoniou and van Harmelen (2004): A Semantic Web Primer. MIT Press. • Fensel et al. (2020): Knowledge Graphs: Methodology, Tools and Selected Use Cases. Springer. • Kerjwal et al. (2021): Knowledge Graphs: Fundamentals, Techniques, and Applications. MIT Press. |
| Methods | <p>The course participants will take part in theoretical and practical exercises, the solutions of which are discussed in the tutorials. At the end of the course, they get the opportunity to apply their knowledge in a team project. Each student team will design and implement a semantic web application, and subsequently present the results to the other students. Besides the exercises, regular presentations including references to relevant course materials and recommended readings will be given by the lecturer. The lecturer as well as the tutors offer individual help and consulting to the participants of the course.</p> |
| Form of assessment | Written examination |
| Admission requirements for assessment | Project report and oral presentation |
| Duration of assessment | 60 minutes |
| Language | English |
| Offering | Fall semester |
| Lecturer | Dr. Sven Hertling |
| Person in charge | Prof. Dr. Heiko Paulheim |
| Duration of module | 1 semester |
| Further modules | - |

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| Range of application | M.Sc. Wirtschaftsinformatik, M.Sc. Mannheim Master in Data Science, M.Sc Mannheim Master in Social Data Science, Lehramt Informatik |
| Semester | 1 st /2 nd /3 rd semester |

D. Data Analytics Methods

| IE 695 | Reinforcement Learning |
|----------------|---|
| Form of module | Lecture with Exercise (partially online) |
| Type of module | Data Analytics Methods |
| Level | Master |
| ECTS | 6 |
| Workload | Hours per semester present: 56 (4 SWS) |
| | Self-study: 98h (70h lectures/exercises, 28h exam preparation) |
| Prerequisites | Machine Learning / Computer Vision /Generative Computer Vision Models course, theoretical and practical knowledge of neural networks |
| Aim of module | <ul style="list-style-type: none"> - Basic concepts of reinforcement learning: MDP, policies, on-policy, off-policy learning - Classical tabular reinforcement learning, DP, Policy Iteration, Q-Learning, SARSA, Monte-Carlo methods - Function approximation for reinforcement learning - Policy gradient methods |

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| Learning outcomes and qualification goals | <p>Expertise:</p> <p>After the course students will understand classical concepts of reinforcement learning as well as state of the art algorithms.</p> |
| | <p>Methodological competence:</p> <p>The students are able to understand and customize popular reinforcement learning algorithms, choose the right setting for their problem and train agents to perform well in environments with which they interact.</p> |
| | <p>Personal competence:</p> <p>The course trains abstract thinking and the ability to formally model application scenarios. By solving assigned exercises independently, the transfer of the learned material to related questions is promoted.</p> |
| Media | Exercise sheets and lecture slides available online, blackboard |
| Literature | - Sutton & Barto: Reinforcement learning: an introduction, 2018 |
| Methods | Lecture, exercises every two weeks, book studies |
| Form of assessment | Written or oral examination |
| Admission requirements for assessment | - |
| Duration of assessment | Written: 90 min. Oral: 25 min. |
| Language | English |
| Offering | HWS |
| Lecturer | Prof. Dr.-Ing. Margret Keuper |

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| Person in charge | Prof. Dr.-Ing. Margret Keuper |
| Duration of module | 1 Semester |
| Further modules | |
| Range of application | Msc Business Informatics, Msc Data Science, Lehramt Informatik |
| Semester | 1./2./3. Semester |

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| DS 204 | Mathematical Foundations of Machine Learning for Social Data Scientists |
| Form of module | Lecture with Tutorial |
| Type of module | Data Analytics Methods |
| Level | Master |
| ECTS | 6 |
| Workload | In presence: 56 h (4 SWS) |
| | Exercises and self-study: 98 h |
| Prerequisites | This is an introductory course and has no pre-requisites. Data Science 100 or 200 is recommended. |
| Aim of module | The aim of the module is to equip students with the mathematics necessary to understand and work with machine learning models. This is an introductory course, pre-supposing only first year college level mathematics, at the level of basic algebra, calculus, and geometry. The module is aimed to get students from their current mathematical location to one where they can understand the basics of mathematics for machine learning, as well as allow them to study more advanced topics with a mathematical dimension. We will cover the material in the first seven chapters of Mathematics for Machine Learning, with a focus on depth of understanding rather than covering a broad range of topics. |
| Learning outcomes and qualification goals | Expertise: The student will gain proficiency in linear algebra, probability theory, and optimization. |

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| | (MK1) |
| | <p>Methodological Competence: The student will be able to combine the methods learned in the class with models encountered in applied questions and work in the social sciences.</p> <p>(MF2, MK3, MF5)</p> |
| | <p>Personal Competence: The student will learn basic proof structures and argument structures that can be used to strengthen their theses and professional arguments.</p> <p>(MK02)</p> |
| Media | Slides will be prepared along with lecture notes; students will collaborate on a set of shared solutions. |
| Literature | Mathematics for Machine Learning. (2020). Marc Peter Deisenroth, A. Aldo Faisal, and Cheng Soon Ong. Cambridge University Press. |
| Methods | The course will involve a lecture where material is discussed as well as tutorial where material is practiced. |
| Form of assessment | Written Examination |
| Admission requirements for assessment | There will be assessments through the semester that will need to be passed in order to take the exam. These may include assessments in lecture or tutorial, as well as take home assignments. |
| Duration of assessment | 90 minutes |
| Language | English |
| Offering | Spring Semester |
| Lecturer | Prof. Dr. Marc Ratkovic |
| Person in charge | Prof. Dr. Marc Ratkovic |
| Duration of module | 1 semester |
| Further modules | - |
| Range of application | M.Sc. Mannheim Master in Data Science (PO 2024), M.Sc. Mannheim Master in Social Data Science |
| Semester | First/Second Semester |

G. Projects and Seminars

| MAS 515 (SEM 477) | Fortgeschrittenenseminar Mathematische Optimierung |
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| Form der Veranstaltung | Seminar |
| Typ der Veranstaltung | Vertiefung |
| Modulniveau | Master |
| ECTS | 4 |
| Arbeitsaufwand | Präsenzstudium: 28 h pro Semester (2 SWS) |
| | Eigenstudium: <ul style="list-style-type: none"> • Vorbereitung des Vortrags: 62 h • Schriftliche Ausarbeitung des Vortrags: 30 h |
| Vorausgesetzte Kenntnisse | Numerik, Optimierung |
| Lehrinhalte | Ausgewählte Themen der Optimierung |
| Lern- und Kompetenzziele | Fachkompetenz: Vertiefte Kenntnisse in einem Spezialgebiet der Numerik / Optimierung (MK1, MK2, MF2) |
| | Methodenkompetenz: Fähigkeit, in einem Spezialgebiet einschlägige Fachliteratur lesen und präsentieren zu können (MF1, MO1, MO3, MO4) |
| | Personale Kompetenz: Kommunikationsfähigkeit (MO3, MO4) |
| Medienformen | Tafelanschriebe, Präsentationen mit Beamer |
| Begleitende Literatur | Ausgewählte Buchkapitel, Zeitschriftenartikel der Numerik / Optimierung / Stochastischen Optimierung |
| Lehr- und Lernmethoden | Vorträge der teilnehmenden Studierenden |
| Art der Prüfungsleistung | Vortrag und schriftliche Ausarbeitung |
| Prüfungsvorleistung | - |
| Prüfungsdauer | - |
| Sprache | Deutsch, auf Wunsch Englisch |
| Lehrende/r | Prof. Dr. Mathias Staudigl, Prof. Dr. Simon Weißmann |

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| Modulverantwortlicher | Prof. Dr. Mathias Staudigl, Prof. Dr. Simon Weißmann |
| Dauer des Moduls | 1 Semester |
| Weiterführende Module | |
| Verwendbarkeit | M.Sc. Wirtschaftsmathematik, M.Sc. Mathematik, M.Sc. Mannheim Master in Data Science |
| Einordnung in Fachsemester | Ab dem 1. Fachsemester |