The Department of Data Science and Knowledge Engineering (DKE) broadly covers Knowledge Engineering in research and teaching.
Dear students,

Welcome to academic year 2020-2021!

About the student handbook
Like every year, the student handbook is here to support you as a source of practical details related to your programme of study. Whether this is your first year with us or whether you are a seasoned student, you should be able to find the answers to many of your study-related questions in this document. To help you find your way around the handbook and our department, we have prepared 8 tips for a flawless time at the Department of Data Science and Knowledge Engineering (DKE). In no particular order, because they are all important in their own right:

• Electronic mail (“e-mail”) is an invention from the 1970s. Its popularity remains high, especially among staff, meaning we will send important information to your student inbox and expect you to read it. You should also keep a close eye on the Student Portal and its announcements section.

• Thank us later, because this tip applies to everything under the sun: whenever you sign up for something, train yourself to read the fine print. Our fine print can be found in something called the ‘EER’ or ‘Education and Examination Regulations’, which outlines the duties and framework for all of us at DKE. You can find the EER in this handbook as well.

• Did you read the EER? Great! Here’s the kind of stuff you can find in there:
  - Are you a new bachelor’s student? As a bachelor’s student, you need at least 45 ECTS (‘credits’, out of a total of 60) in the first year in order to continue with the programme.
  - Second year bachelor’s students need the full 60 ECTS from year 1 and at least 40 ECTS from year 2 to participate in the year 3 curriculum.
  - First year master’s students need to fulfill 40 ECTS from year 1 to participate in the year 2 curriculum.

• Ah, the perks of being a DKE student – a subsidized MSV Incognito membership, an annual free barbecue, daily mathematics and the freedom (ok, obligation) to book your own courses. Freedoming is difficult and that is why we allow you to forget booking once, but only once, per year.

• The first rule of DKE is: you do not plagiarize. The second rule of DKE is: you do not plagiarize. The third rule of DKE: if you do not know what we mean by plagiarism, please have a look at article 5.15 of the EER.

• While at the coffee machine, try not to mistake a lecturer for a student (unless you turn it into a polite comment about how young they look): memorizing staff’s faces is already possible using the last pages of this handbook.
• Speaking of coffee machines: we do have a code of conduct – to be found in this handbook as well. If you are not the type to eat pizza with your feet on the table during a lecture, common sense may already get you pretty far (again, this applies to life in general and at DKE).

• Imagine: you are about to graduate... Or so you think, until you hear that you accidentally booked courses that do not match the graduation requirements. Please save us both from this terrible situation. Master’s students will want to read subsection 1.2b to brush up on the possibilities and restrictions of course booking.

Wishing you a wonderful year,

Pietro Bonizzi  
Director of Studies bachelor’s programme

Mark Winands  
Director of Studies master’s programmes
Research at DKE spans the disciplines and interfaces of artificial intelligence, data science, computer science and applied mathematics.
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Project-Centred Learning (PCL) is a teaching method used in all of the Department of Data Science and Knowledge Engineering’s programmes.
1 Education: The Profile of the Study of Data Science and Artificial Intelligence

1.1 What is the programme about?

Knowledge is the central factor in modern society. Smart chips help companies to keep track of goods and to manage supplies and stocks. The usability of new high-tech communication electronics, such as smartphones, navigation devices and digital cameras, is greatly enhanced by intelligent software. Electronic banking and e-commerce heavily rely on real-time access to large databases and require highly safe protocols for information exchange. Medical and biological engineering helps medical doctors to arrive quickly at an accurate diagnosis, and is now opening up exciting new possibilities for personalized medicine. In the Data Science and Artificial Intelligence programme, you learn how to approach new challenges in these important areas. You learn to collect and organize valuable information, with the help of mathematics and modern computer techniques. You learn to use models and computers to analyse the information and to generate new knowledge, to draw important conclusions, to solve practical problems efficiently, and to speed up the decision-making process.

The Data Science and Artificial Intelligence programme offers you a unique combination of courses in applied mathematics, computer science and artificial intelligence. From the computer science point of view, the emphasis is on software, programming, algorithms and logic. Courses in applied mathematics are designed so that you quickly become acquainted with the important concepts, methods and techniques - always from a practical point of view, as proofs are never taught for mathematics sake only. Artificial intelligence provides you with ways to reason with available knowledge, and it introduces you into the world of machine learning and intelligent search. Did you know that, nowadays, artificial intelligence is a standard component of commercial computer games?
What makes the course programme stand out from traditional educational programmes in computer science or mathematics, is the goal-oriented, practical and applied character. Mathematical and computer models continuously serve this goal, by bridging the gap between theory and practice. Classic optimization methods and modern intelligent techniques allow you to arrive at efficient and sometimes surprisingly elegant solutions. They help you to value the methods and techniques you encounter, and to gain insight into their usefulness for dealing with practical problems.

This unique profile of Data Science and Artificial Intelligence is the key to a flexible educational programme that prepares you for a successful professional career in modern society. The wide range of application areas and the project-centred educational model, offers you the opportunity to experience a diversity of topics from the ICT and

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*The bachelor programme will be renamed as “Data Science and Artificial Intelligence” in 2020–2021.*
telecommunications industry, medicine and biology, science, economics and business management, just to name a few. Optional courses, your choice of topic for the bachelor’s thesis, internships, and the possibility to study abroad, allow you to tune the contents of your study to your personal interests.

1.2 Study System

1.2.1 The bachelor’s programme curriculum

The bachelor’s programme in Data Science and Artificial Intelligence is a three-year programme. We chose for a broad setup of the curriculum, so that students can decide on the way they would like to specialize during the final stage.

### Year 1

<table>
<thead>
<tr>
<th>Period 1</th>
<th>Period 2</th>
<th>Period 3</th>
<th>Period 4</th>
<th>Period 5</th>
<th>Period 6</th>
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<tbody>
<tr>
<td>Introduction to Computer Science 1; Discrete Mathematics; Introduction to Data Science and Artificial Intelligence</td>
<td>Introduction to Computer Science 2; Linear Algebra; Computational and Cognitive Neuroscience</td>
<td>P</td>
<td>Data Structures and Algorithms; Logic; Numerical Mathematics; Software Engineering</td>
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<td>Calculator; ICT and Knowledge Management</td>
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### Year 2

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<tr>
<th>Period 1</th>
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<th>Period 3</th>
<th>Period 4</th>
<th>Period 5</th>
<th>Period 6</th>
</tr>
</thead>
<tbody>
<tr>
<td>Databases; Probability and Statistics; Graph Theory</td>
<td>Machine Learning; Linear Programming; Reasoning Techniques</td>
<td>P</td>
<td>Theoretical Computer Science; Mathematical Modelling; Human Computer Interaction and Affective Computing</td>
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### Year 3

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<th>Period 1</th>
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<th>Period 3</th>
<th>Period 4</th>
<th>Period 5</th>
<th>Period 6</th>
</tr>
</thead>
<tbody>
<tr>
<td>Semantic Web; Game Theory; Prolog; Computer Security; Robotics and Embedded Systems; Digital Society</td>
<td>Large Scale IT and Cloud Computing; Logic for AI; Parallel Programming; Introduction to Bio-Informatics; Software and Systems Verification; Quantum Computing</td>
<td>P</td>
<td>Data Analysis; Operations Research Case Studies; Intelligent Systems</td>
<td>R</td>
<td>E</td>
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<tr>
<td>PROJECT</td>
<td>PROJECT</td>
<td></td>
<td>BACHELOR 'S THESIS</td>
<td>BACHELOR'S THESIS</td>
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</tr>
</tbody>
</table>
*Third year students choose 6 optional courses (3 per period) in addition to the semester project in semester 1 of year 3. In case students have passed both electives of period 2.5, either the course Natural Language Processing or Introduction to Image & Video Processing can replace 1 of the third year electives. Student can also choose (1) elective courses at other UM bachelor programmes of at most 18 ECTS (2) the minor Entrepreneurship or (3) the educational minor. (4) In addition, students can study abroad for a semester at one of our exchange partners (see Section 1.4). Please contact the study advisor for more information. Also, check the Study Abroad section on through the “My Organisations” section of the Student Portal.

Periods 1 and 2 last eight weeks in total. During week 1-7 there are classes and in week 8 exams. Periods 4 and 5 will last 9 weeks due to several holidays in this period. Three courses are offered during each period, each course is good for 4 credits (ECTS). Per course, five to seven hours of class are offered each week in year 1, and five hours in year 2 and 3. Note that year 1 has an attendance requirement for resitting an Exam: check the Education and Examination Regulations (EER) that are published in this study guide and through the “My Organisations” section of the Student Portal for more information.

Next to these courses, you participate in a project of 6 credits that will last the whole semester. Skill trainings and project meetings are mandatory: you are expected to be present during 100% of the skill classes and 100% of the project meetings in each academic year. If a student fails to do so, the project examiners may deviate from the group grade for this individual student. The project of semester 1 runs during period 1, 2 and 3. The project of semester 2 runs during period 4, 5 and 6. For specific details on the project curriculum in year 1 see section Project 1-1 and Project 1-2 with the course descriptions of year 1.

Periods 3 and 6 will last three weeks, during which students work full-time to finish their project assignment. After the project weeks, it is possible to resit previous periods during the fourth week. During both project periods, you will work full-time on a project assignment. This project assignment is announced in the beginning of periods 1 or 4, along with the group composition. At the end of week 7 of the 8-weeks periods, each group separately gives a brief presentation for the teachers of the subjects concerned. Each group shows their interpretation of the problem and will discuss their approach and schedule for the further elaboration. The teachers give their reaction and judge whether the subjects are sufficiently covered in the project. These interim presentations will be assessed. The elaboration stage of the project weeks is concluded with the handing in of the final report, handing in of the product, and giving an oral presentation of the project results for all groups. The assessment of the report, the presentation, and the product in principle result in the same mark for all the group members.

The final stage of your bachelor’s programme, period 5 and 6 of year 3 is reserved for writing your bachelor’s thesis that equals 18 credits. Every student has to conduct a short scientific research focussed on a relevant topic. This can be empirical or theoretical research. Students have acquired information on these different research domains throughout their educational programme. Each student has to hand in a signed bachelor’s project plan to the Bachelor’s thesis coordinator. After approval, students are placed in a group, based on their topics. A thesis supervisor with experience in the field of choice is appointed to each student. In the second period of this semester, the students conduct their own research. In the end of the last period of the semester, each student has to present his or her results.
DKE Honours Programme
The DKE Honours Programme consists of Knowledge Engineering@Work (KE@Work) and (MaRBLe 2.0). Students admitted to the KE@Work path are placed at a company or organization in the region through a careful selection and matching process. During the full second and third year of the bachelor’s programme, they spend 50% of the time in class and 50% at the company, where they work on solving academic challenges and complex business problems, under supervision of dedicated business and DKE supervisors. During MaRBLe 2.0 you will get the opportunity to work on a state-of-the-art research project. Work will be organized in a similar way as in professional research institutes where participants work together as individual experts on a team project. Participation is open to excellent and motivated students.

1.2b The master’s programme curriculum
We offer two master’s programmes that are both two-year, programmes conducted in English: Artificial Intelligence (AI) and Data Science for Decision Making (DSDM). While the programmes have been designed with a September-start in mind, you can start either in September or in February (please check the website for admission deadlines). The tables below show the curriculum of AI and DSDM. When you enrol in September, you will start in semester 1, and when you enrol in February, you will start in semester 2. Mandatory (core) courses are underlined.

Master’s AI Year 1

<table>
<thead>
<tr>
<th>Semester 1 (September intake)</th>
<th>Semester 2 (February intake)</th>
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<tbody>
<tr>
<td>Period 1</td>
<td>Period 2</td>
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<tr>
<td>Foundation of Agents;</td>
<td>Multi-Agent Systems;</td>
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<td>Intelligent Search</td>
<td>Advanced Concepts</td>
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<tr>
<td>and Games</td>
<td>in Machine Learning</td>
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<td>PROJECT</td>
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Year 2

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<thead>
<tr>
<th>Semester 3</th>
<th>Semester 4</th>
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<tbody>
<tr>
<td>Period 1, 2 and 3</td>
<td>Period 4, 5 and 6</td>
</tr>
<tr>
<td>Elective Semester: Courses; Research Project/Internship; Business Internship; Study Abroad</td>
<td>THESIS</td>
</tr>
</tbody>
</table>
Master Data Science for Decision Making Year 1

<table>
<thead>
<tr>
<th>Period 1</th>
<th>Period 2</th>
<th>Period 3</th>
<th>Period 4</th>
<th>Period 5</th>
<th>Period 6</th>
</tr>
</thead>
<tbody>
<tr>
<td>Data Mining; 1 of the following: Signal and Image Processing; Mathematical optimization; Stochastic Decision-making</td>
<td>Model Identification and Data Fitting; 1 of the following: Advanced Concepts in Machine Learning; Applications of Image and Video Processing; Information Security</td>
<td>PROJECT</td>
<td>Algorithms for Big Data; 1 of the following: Dynamic Game Theory; Computational Statistics Building and Mining Knowledge Graphs Advanced Natural Language Processing</td>
<td>Planning and Scheduling</td>
<td>PROJECT</td>
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<td>PROJECT</td>
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Year 2

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<tr>
<th>Semester 3</th>
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<tbody>
<tr>
<td>Period 1, 2 and 3</td>
<td>Period 4, 5 and 6</td>
</tr>
<tr>
<td>Elective Semester: Courses; Research Project/Internship; Business Internship; Study Abroad</td>
<td>THESIS</td>
</tr>
</tbody>
</table>

Periods 1 and 2 will last eight weeks in total. During week 1-7 there are classes and in week 8 exams. Periods 4 and 5 will last 9 weeks due to several holidays in this period. Each course equals 6 credits (ECTS). Per course, five hours of class are offered per week, in which a teacher will explain the theory of the subject or in which you have to do some practical training.

Depending on the master’s programme you enrolled in, you are required to pass a number of mandatory courses (5 for the AI master, and 4 for the DSDM master). These courses are underlined in the tables above. Besides these courses, you can choose particular electives to complete your first-year curriculum. The lectures of each course will be scheduled, so that there will be no overlap between classes. The days that you will have class will depend on your course selection.

Next to these courses, you participate in a project of 6 credits that will last the whole semester. As a student, you are expected to participate actively in doing tasks with respect to the project skills training and project meetings. In addition, students are expected to cooperate actively with their group in order to successfully finish your project assignment. If a student fails to do so, the project examiners may deviate from the group grade for this individual student. The project of semester 1 runs during period 1, 2 and 3. The project of semester 2 runs during period 4, 5 and 6. Periods 3 and 6 will
last four weeks, during which students work fulltime to finish their project assignment in the first 3 weeks and take possible re-examinations of the previous periods during the fourth week. During both project periods, you will work full-time on a project assignment. This project assignment will be announced in the beginning of periods 1 or 4, and the group composition and project topic assignment is based on student preference in so far as possible. At the end of week 7 of the 8-weeks periods, each group separately will give a short presentation. Each group will describe their project tasks and discuss their approach and schedule for the further elaboration. The examiners will give their reaction and judge whether the subjects are sufficiently covered in the project. These interim presentations are assessed. The elaboration stage of the project weeks will be concluded with the handing in of the final report, a dissemination website, handing in of the product, and a public oral presentation of the project result for each group. The assessment of the report, website, presentation and the product will - in principle – result in the same mark for all the group members.

Choose your own curriculum
During the first semester of the second year of the master’s programme, you can choose your own curriculum, enabling you to pursue your personal interests. During this semester, you can obtain 30 ECTS by choosing elective courses of the other master’s programmes offered at DKE (i.e., AI or DSDM). In addition, you can also:
• take a combination of elective courses at DKE and another faculty of Maastricht University;
• participate in a research project of the academic staff or at a another university;
• participate in an internship at a company;
• follow an exchange programme at one of our partner universities abroad;
• participate in an honours programme offered by Maastricht University (extracurricular).

The final stage of your master’s programme, during the second semester of the second year, is reserved for writing your master’s thesis that counts for 30 ECTS. The thesis is produced individually and is the result of a master research project on a topic that you will be working on under the supervision of one of the academic staff members. In the first phase, the emphasis is on self-study, subject determination, approaching a supervisor, planning and some preliminary research. After approval of the thesis research plan by the Board of Examiners, the actual research is started in the second phase. In this phase, the student carries out his/her own research. The senior researcher that acts as the supervisor of this research process will guide the student using a series of frequent appointments. The final phase is used to accomplish, i.e. write, the master’s thesis. The master’s thesis project is completed by an individual presentation of the results. Assessment will be based on the research, the thesis itself, the process, the software and the presentation of this thesis (i.e., public defence).

Note that all individual curriculum choices are guided by our study advisor and academic staff and will always be evaluated on quality by the Board of Examiners.
1.3 Project Centred Learning

The programmes of the Department of Data Science and Knowledge Engineering are designed around the Project-Centred Learning (PCL) teaching method. It resembles the Problem Based Learning style, for which Maastricht is well known. The PCL educational model is small-scale and student-oriented. You work in small groups on complex and challenging projects that require you to develop a variety of skills. Companies and institutes – who often submit projects - give our students the opportunity to gain invaluable experience by applying their education to finding solutions to real-world problems. Therefore, you immediately apply what you have learned from the course material and lectures to real-life problems. Together with fellow students, you research which information is required and how it is best presented. At the end of each project, you deliver a functional computer programme and present your findings to your fellow students, the teachers and/or the client.

Project-Centred Learning has the following advantages:
• from the beginning you find out what teamwork means
• you learn project-related skills in a natural way
• you will be continuously placed in an active role
• you will be able to match theory with its applications
• PCL increases the student’s motivation

Some examples of Master’s projects conducted at DKE:
• Finding “Banksy” through Image Processing
• Automatic Generation of Contextual Celtic Knotwork
• Modelling Human Decision Process from Intracranial EEG
• Relating component responses between rats and humans
• Kick-optimization for Robotic Soccer

1.4 Internationalization

The Department of Data Science and Knowledge Engineering has one of the highest ratios of students from abroad. More than 70% of the scientific staff and 70% of the students are non-Dutch, giving rise to an international study environment. Additionally, DKE hosts a number of international exchange students each year and offers its students a number of opportunities for international experiences themselves. For example, DKE offers its students the opportunity to study abroad for a semester during the elective semester. To this purpose, we collaborate with well-established universities such as:

**European Economic Area (EEA)**
- University of Copenhagen
- Reykjavik University
- Sapienza Università di Roma
- Université Toulouse Capitole °
- Université du Luxembourg °
- Universität Konstanz °
- Aarhus University x
- Université Paris Dauphine °

**NON EEA**
- The University of Sydney
- University of Technology Sydney
- Université de Montréal °
- Université de Genève °
- City University of Hong Kong °
- Singapore Management University °
B2 level French required
* Only Master courses
° Only Bachelor courses

The above-mentioned exchange partners are an indication and subject to change due to (temporary) agreements and exchange student balance. For an overview of the available partners per semester, see the ‘exchange application’ web form.

Bachelor’s students can participate in an international study abroad programme at a selection of these universities during the first semester of the third year of the bachelor’s programme. Master’s students at a selection of these universities during the first semester of the second year of the master’s programmes.

Please note that for students entering the Master programmes during the February Intake, fewer universities will be available. Additional international collaboration is done through the exchange of course coordinators with RWTH Aachen, with which we closely collaborate and the fact that our two master’s programmes Artificial Intelligence and Data Science for Decision Making are embedded in the School of Information Technology (SIT) of the transnational University Limburg (tUL). For more information, please contact the study advisor and read the Study Abroad Guide available on through the “My Organisations” section of the Student Portal.

1.5 Degree

A successful conclusion of the bachelor’s programme will provide you with a bachelor’s certificate according to Dutch law, that is, a ‘Bachelor of Science’.
A successful conclusion of a master’s programme will provide you with a master’s certificate according to Dutch law, that is, a ‘Master of Science’.

1.6 Study programme’s feasibility and Quality Assurance

DKE strives for continuous improvement of the quality and feasibility of its study programmes. Student evaluations of each course help us in maintaining a high standard of educational quality and keeping the study programmes feasible. The study programme’s feasibility means that a student with an appropriate background should be able to finish the study within the set number of years.
To maintain this standard, the quality assurance officer collects information about teaching, learning and assessment at the end of each period. The quality assurance officer then reports the outcomes of the student evaluations to the DKE Education Programme Committee (EPC), which also includes four student-representatives.
If the outcomes are unsatisfactory, the EPC will take action to improve the quality of a specific course (or project) or of the study as a whole. Therefore, student responses are essential in pointing out strong aspects and aspects for improvement of all educational activities.

As a DKE student, you are encouraged to give your feedback on each course, as it may give cause for adjustments of that course. Moreover, future students may benefit from the results and comments of your evaluation just as you may benefit from course evaluations of fellow students.
2 Curriculum of the Bachelor’s programme and Master’s programmes

2.1 Curriculum of the First Year of the Bachelor’s Programme

In order to learn how the processing and preparation of knowledge is performed with the help of computer systems, a thorough basic knowledge of specific mathematics and computer science subjects is required. This means that the first year is largely filled with mathematics and computer science subjects. Apart from that, you will also get introduction to Computational & Cognitive Neuroscience and ICT & Knowledge Management. The year is divided into four periods of eight weeks with three courses each, and two periods of four weeks during which you will work on a project. Each project is preceded by partial project assignments during the other periods. The week schedule works with two-hour clusters. In the overview below, the courses are indicated, as well as the study load in credits (ECTS). One ECTS stands for about 28 hours of study time (lectures, meetings and self-study). Besides the lectures that are given on the subjects, there will also be practicals and skills training.

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<tr>
<th>Year 1</th>
<th>ECTS</th>
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<tbody>
<tr>
<td>Period 1.1</td>
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<tr>
<td>Introduction to Data Science and Artificial Intelligence (KEN1110)</td>
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<tr>
<td>Introduction to Computer Science 1 (KEN1120)</td>
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<tr>
<td>Discrete Mathematics (KEN1130)</td>
<td>4</td>
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<tr>
<td>Project 1-1 (*)</td>
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<tr>
<td>Period 1.2</td>
<td></td>
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<tr>
<td>Computational &amp; Cognitive Neuroscience (KEN1210)</td>
<td>4</td>
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<tr>
<td>Introduction to Computer Science 2 (KEN1220)</td>
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<tr>
<td>Linear Algebra (KEN1410)</td>
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<tr>
<td>Project 1-1 (*)</td>
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<td>Period 1.3</td>
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<td>Project 1-1 (KEN1300)</td>
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<td>Period 1.4</td>
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<tr>
<td>Calculus (KEN1440)</td>
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(*) Project 1-1 will start in period 1.1; Project 1-2 will start in period 1.4. The credits for the projects will become available at the end of period 1.3 and period 1.6, respectively. Please see the course description section Project 1-1 and Project 1-2 for more details on the project curriculum. For each period, we will give a short explanation of the various parts. Before the start of each period, the students will receive detailed information about the content, the study material, the teaching form, the schedule, and the examination method.
Period 1.1

Introduction to Computer Science 1 (Code: KEN1120)
Examiner: Dr. G. Spanakis and Dr. E. Hortal Quesada
Desired Prior Knowledge: None. The course appears as desired prior knowledge for the courses Introduction to Computer Science 2, Data Structures and Algorithms, Software Engineering, Databases and Machine Learning.
Prerequisites: None. It appears as part of the pre-requisites of the second semester project in year 1, both projects of year 2 and the third year courses Prolog, Parallel Programming and Robotics.
Description: The course provides the basics of computer science and computer programming. After a short introduction to computer organization, the principles of programming are presented. The main topics of the course are: data types, variables, methods, parameters, decision structures, iteration, arrays, recursion and a brief introduction of objects. Programming skills will be acquired during practical sessions using the object-oriented programming language Java.
Knowledge and understanding: The course offers preliminary methodological and theoretical bases for studying and applying computers and computer programming on which the rest of the curriculum builds.
Applying knowledge and understanding: Whenever a computer system or a programming system has to be designed and implemented the knowledge and insights acquired during the course can be used and applied.
Making judgements: After successful completion of the course, students will be able to judge the quality and correctness of simple non-object-oriented programs.
Communication: The skills acquired during the course will enable student to communicate about standard programming constructs and algorithmic basics.
Learning skills: After successful completion of the course students will be able to formalize, analyse and program solutions to simple software problems.
Exam: Closed-book written exam (80%) + Assignments (20%)
ECTS: 4

Discrete Mathematics (Code: KEN1130)
Examiner: Dr. S. Kelk and Dr. M. Musegaas
Tutor: Dr. J. Kuipers
Desired Prior Knowledge: None.
Prerequisites: None.
Description: In this course, we build a mathematical framework that is based on logic and reason. The main objective of the course is to make students familiar with the language of mathematics. Students will learn how to make sound arguments and to detect where and why certain arguments go wrong. For this purpose, we will discuss the basic principles of logic and, closely related, the basic types of mathematical proofs. In doing so, we will encounter numbers such as integers, natural numbers and real numbers and we shall examine what makes these numbers special. After that, we will use basic logic to discuss, among other things, the following mathematical concepts: infinity, sets, relations, functions, permutations and combinations. Our fundamental tool in all of this is plain common sense. You really do not need your toolbox of mathematical formulas learned in previous studies and neither do you need a calculator. Pen and paper are the basic instruments needed. After completing each
topic, exercises will be provided to be completed in class or at home, since mathematics is mainly learned by practising repeatedly.

**Knowledge and understanding:** Students will be able to read, interpret and manipulate basic mathematical terminology (propositional logic, quantifiers, set theory, relations, functions, and combinatorics). Students will also be able to read and interpret several different types of mathematical proofs and identify whether a purported proof is mathematically sound.

**Applying knowledge and understanding:** Upon completion of the course students will know how to read, interpret, write and manipulate rigorous mathematical statements using propositional logic, quantifiers, set theory, relations, functions and combinatorics. Students will be able to select, from a range of mathematical tools, which is appropriate to prove or disprove a given mathematical statement, and apply the chosen tools, rigorously and clearly in order to achieve the desired goal.

**Making judgements:** Students will be able to distinguish between mathematically sound and unsound statements and defend the rigour of their own mathematical arguments.

**Communication:** Students will be able to write clear, rigorous and explicit mathematical arguments using standardized mathematical terminology and such that each step in the argument is a logical consequence of earlier steps.

**Learning skills:** By the end of the course, students will be able to autonomously and critically reflect upon the mathematical correctness of their own arguments.


**Recommended literature:** None

**Exam:** Written exam and optional weekly bonus exercises (the results of which are added to your exam score, up to 10%).

**ECTS:** 4

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**Introduction to Data Science and Artificial Intelligence (Code: KEN1110)**

**Examiner:** Dr. P. Bonizzi, Dr. R. Cavill and Dr. A. Wilbik

**Desired Prior Knowledge:** None.

The course appears as desired prior knowledge for the courses Reasoning Techniques and Theoretical Computer Science.

**Prerequisites:** None.

**Description:** The course Introduction to Data Science and Artificial Intelligence offers a comprehensive overview of the core topics in Data Science and Artificial Intelligence (DKE), both from a mathematical and from a computational perspective. Particular emphasis will be put on the basic classes of techniques and methods, the theoretical underpinnings of data science and computational intelligence, and some example application domains of data science. As such, the course will provide an overview of many topics that are addressed in much more detail throughout the Bachelor’s programme.

**Knowledge and understanding:** The aim of this course is to make students recognise what real world problems require the use of data science, and approach their solution by using a data science process, namely: explore the data, model the data, and perform simulations if required. Moreover, they will exhibit knowledge in the basic concepts of artificial intelligence, such as agents, search, artificial intelligence, decision trees.

**Applying knowledge and understanding:** Students learn to recognise applications of data science and knowledge engineering in different domains and apply the basic techniques they have learnt from both.

**Making judgements:** Upon completion of the course, students are able to recognise the relevant domains of data science and artificial intelligence when confronted with DKE-problems.
Communication: Students are able to explain the process they used to generate results and communicate the meaning of those results in context.

Learning skills: Students have acquired the skills to recognise small-scale data science problems and autonomously and critically reflect upon the appropriateness of the data science process for tackling those, and propose a primary solution.

Study material: Material will be provided during the course.

Recommended literature:

Exam: There will be a closed book written exam at the end of the course.

ECTS: 4

Period 1.2

Introduction to Computer Science 2 (Code: KEN1220)
Examiner: Dr. E.N. Smirnov and Dr. E. Hortal
Desired Prior Knowledge: Introduction to Computer Science 1
Prerequisites: None.

Description: This course is a follow-up of the course Introduction to Computer Science 1. It teaches object-oriented programming in Java. The main topics covered in the course are objects and classes, interfaces and polymorphism, event handling, inheritance, graphic user interfaces, exception handling, and streams.

Knowledge and understanding: After successful completion of the course, students will be able to explain the methodological and theoretical principles of object-oriented programming.

Applying knowledge and understanding: Students will be able to implement basic object-oriented computer programs. They will be able to design and describe simple object-oriented computer systems.

Making judgements: Students will be able to judge the quality and correctness of simple object-oriented programs.

Communication: Students will be able to communicate about object-oriented programming constructs and algorithmic basics.

Learning skills: Students will be able to recognize their own lack of knowledge and understanding and take appropriate action such as consulting additional material or other sources of help.

Course notes, slides, and other information made available.


Exam: Written exam (80%) + practical assignments (20%).

ECTS: 4
Linear Algebra (Code: KEN1410)
Examiner: Dr. S. Chaplick and Dr. S. Mehrkanoon
Tutor(s): None.
Desired Prior Knowledge: None. The course itself occurs as part of the pre-requisites of the second semester project in year 1, and as desired prior knowledge for the second year courses, Mathematical Modeling, Linear Programming.
Prerequisites: None.
Description: This course introduces the fundamental concepts of linear algebra, and examines them from both an algebraic and a geometric point of view. First, we address what can be recognized without doubt as the most frequently occurring mathematical problem in practical applications: how to solve a system of linear equations. Then we discuss linear functions and mappings, which can be studied naturally from a geometric point of view. Vectors spaces are then introduced as a common framework that brings all themes together. Next, we shift from the geometric point of view to the dynamic perspective, where the focus is on the effects of iterations (i.e., the repeated application of a linear mapping). This involves a basic theory of eigenvalues and eigenvectors, which have many applications in various branches of science as for instance in problems involving dynamics and stability, in control theory, and in optimization problems found in data science. Key concepts in the course are: vectors, matrices, systems of linear equations, eigenvalues, eigenvectors, linear transformations, and orthogonality. The software package Matlab is introduced in the accompanying computer classes, where emphasis is put on the application of linear algebra to solve real world problems.
Knowledge and understanding: Students are able to recognize and explain the fundamental concepts of Linear Algebra: systems of linear equations, vectors and vector spaces, basis and coordinates, matrices and matrix-vector computations, linearity and orthogonality, linear independence, rank, fundamental spaces (row space, column space, null space), determinants and invertibility, eigenvalues and eigenspaces, diagonalization.
Applying knowledge and understanding: Students are able to analyse a linear algebra problem from both an algebraic and a geometrical point of view. Students can solve systems of linear equations, compute determinants and rank, compute eigenvalues and eigenspaces, make use of complex numbers, diagonalize matrices, perform change of coordinates.
Making judgements: Students are able to look at the same problem from different angles and to switch their point of view (from geometric to algebraic and vice versa).
Communication: Students are able to motivate, both from an algebraic and a geometric point of view: the solution set of a system of linear equations, the linear independence and orthogonality of a set of vectors, the linear transformation between two coordinate systems, the fundamental spaces associated with a matrix, the invertibility of a matrix, and the diagonalization of a matrix in terms of the properties of its eigenvalues and eigenvectors.
Learning skills: Students have acquired the skills to autonomously recognize elements of practical problems, which can be addressed and solved with linear algebra, and use Matlab to solve larger scale problems.
Recommended literature: None.
Exam: There will be a closed book written exam at the end of the course.
ECTS: 4
Computational and Cognitive Neuroscience (Code: KEN1210)
Examiner: Dr. A. Roebroeck and Dr. M. Capalbo
Desired Prior Knowledge: None.
Prerequisites: none.
Description: The course Computational and Cognitive Neuroscience presents an overview of the core topics in cognitive and biological psychology. These topics include (human) perception, learning, memory, planning, problem solving, reasoning, language, speech, and action. Both the functional and neuroanatomical foundations of cognitive faculties are addressed. Several models of cognition and theories of brain function that are of relevance to knowledge engineering will be outlined. Several skills trainings will be given to train understanding in biological functioning of neuronal communication, and functioning of neural networks and genetic algorithms.
Knowledge and understanding:
The student can recount the main points of the domain of cognitive science
• The student can describe the main points of the domain of cognitive science
• The student can explain the following (human) behaviours while using these points: perception, learning, memory, planning, problem solving, reasoning, language, speech, and action.
• The student can identify the computational aspects and computational applications of these fields
Applying knowledge and understanding:
• This knowledge is applied in two practical assignments in which the students are asked to create a genetic algorithm and a neural network
Making judgement:
• Upon completion of the course, students are able to interpret data and literature about a subject in (or related to) the domain of cognitive and biological psychology.
• Using the data and literature, they can support judgements about the societal, scientific or ethical aspects of the subject.
Communication:
• Students are able to communicate ideas and solutions to an audience of non-experts and experts.
Learning skills:
• Students have acquired the skill to translate theoretical models into computational models.
Study material: Material will be provided during the course.
Exam: Written exam
ECTS: 4

Period 1.3

PROJECT 1–1 (Code: KEN1300)
Examiner: Dr. Chaplick, Dr. E. Smirnov, Dr. S. Mehrkanoon, Dr. E. Hortal-Queasada, Dr. M. Staudigl, and Dr. J. Niehues
Coordinator: Dr. K. Schüller
Tutors: Dr. K. Schüller, Dr. Popa, Dr. O. D’Huys
Description: Students work on a project assignment in small groups of about six
students. The group composition stays the same for the whole project and is announced at the beginning of period 1.1. The students are guided through the project by a fixed tutor. The project assignment is divided into three subtasks (one per period) and is strongly related to the content of the courses from period 1.1 and 1.2. In period 1.1, after receiving the assignment for the whole project at the end of week 5, the students work full-time on the project in week 6. In this week, each group meets the tutor twice. In period 1.2, the students continue working on the project, while also having to attend the courses of that period. They meet their tutor approximately once a week. In period 1.3, the students work three weeks full-time on the project and meet their tutor twice a week. At the beginning of period 1.2 and 1.3, the students have to hand in a planning for the current phase. At the end of each period, the students have to give a presentation and the source code, presentation and an overview of who did what need to be uploaded to Student Portal. While the presentations at the end of period 1.1 and 1.2 are in front of the examiners and the tutors, the presentations at the end of period 1.3 will additionally be in front of the fellow students. In period 1.3, they furthermore have to hand in a report and attend a product and report examination. 

**Knowledge and understanding:** Interpret constraint-satisfaction problems arising in practice and translate this to discrete-mathematical algorithmic models capable of solving the problem. Gain insight into practical use of basic software design and development principles. Recognise and relate user-computer interactions to concepts from graphics and user-interface frameworks. Strengthen knowledge of basic algorithms and methods for efficiently solving constraint-satisfaction problems arising in applied mathematics (especially: discrete mathematics) and artificial intelligence.

**Applying knowledge and understanding:** Design an answer strategy for scientific questions using analytical thinking and logical reasoning. Translate discrete-mathematical algorithmic models to software code. Implement software to efficiently solve constraint-satisfaction problems arising in applied mathematics (especially: discrete mathematics) and artificial intelligence by finding, designing and applying appropriate algorithms. Formulate computational experiments, and analyse and interpret the results. Apply basic design and development principles in the construction of software systems. Use existing software application frameworks for graphics and user interfaces. Use tools for software project management such as version control systems and issue trackers. Identify project goals, deliverables, and constraints. Plan and chair meetings. Create notes for minutes. Work in a team such that the workload is balanced. Plan teamwork by setting deadlines and distributing tasks.

**Making judgements:** Evaluate different mathematical and computational models with respect to their suitability, efficiency and correctness for a specific task. Elicit and evaluate relevant scientific background information.

**Communication:** Give a clear and well-constructed presentation, including a demonstration of the product, and with appropriate use of illustrations and/or videos. Offer and respond to questions on and constructive criticism of presentations. Write a project report according to the structure of an academic article. Submit arguments in exact sciences, with appropriate use of formulae and figures. Cite published sources in the project report according to the academic guidelines. Structurally inform stakeholders on project progress. Effectively communicate with project group members about task division, planning and project deadlines. Effectively communicate with group members by listening to others’ ideas; be contactable include others in the discussion. Cooperate in a group to reach a consensus view. Give constructive feedback to team members. Communicate in the English language.

**Learning skills:** Reflect on one’s own academic abilities and functioning in a team.

**Study material:** Project manual project 1-1, Maastricht University, DKE
Assessment: Final grade = 0.15 x grade of phase 1 + 0.15 x grade of phase 2 + 0.7 x grade of phase 3, where grade of phase 3 = 40% grade for product + 40% grade for report + 10% grade presentation + 10% grade project management. Missing project meetings and skill classes will cause an individual grade reduction of the final grade. Furthermore, the examiners may deviate from the group grade for individual students if they either contributed significantly more or less than other group members.

Skill classes:
- **Introduction into PCL (period 1.1)**
  The students learn how DKE applies the PCL approach. The structure of the periods regarding the project is explained. Furthermore, the students will work on a small assignment in groups and reflect on their group work.
- **Team Dynamics 1 (period 1.1)**
  This team dynamics workshop aims to provide you with a deeper awareness, insight and practice in effective team collaboration & co-creation. During this introduction workshop, you and your project team will draw up a team charter (contract) to initiate effective group collaboration in project 1-1.
- **Minutes, agenda and planning (period 1.1)**
  In this skill class, you will learn how to create an agenda, chair a project meeting and to take notes that can be understood by someone not attending the meeting. Furthermore, you will learn how to create a proper planning for a project in form of a Gantt chart.
- **Elementary Writing and Presenting (period 1.2)**
  This skill class is divided into two parts: 1) Elementary Writing: Here, you learn how to structure a report/article and what content to write in each of the sections of a report. You learn how to use signal words, referral words and basics of how to write in the mathematical language. 2) Presenting: In this part, you learn basic knowledge about body language, group presentations and slide layout.
- **Citing and Referencing (period 1.2)**
  During this class, you will learn about when and how you should cite various (scientific) sources in your project report.
- **Team Dynamics 2 (period 1.2)**
  In this skill class, you are going to evaluate the team collaboration and communication during project 1-1 by means of interactive exercises.
- **Presentation Skills (period 1.3)**
  In this skill class, you will present within your project group and receive individual feedback on presentation style and content.

**ECTS: 6**

**Period 1.4**

**Data Structures and Algorithms (Code: KEN1420)**

*Examiner:* Dr. J. Niehues and T. Pepels, M.Sc.

*Tutor(s):* TBA.

*Desired Prior Knowledge:*
Discrete Mathematics, Introduction to Computer Science 1 and 2. The course is desired prior knowledge for Theoretical Computer Science.

*Prerequisites:*
None. The course itself occurs as part of the pre-requisites of both projects of year 2 and the third year course Parallel Programming.

*Description:*
As a continuation of the courses Computer Science 1 and 2, this course will treat the systematic design and application of data structures and algorithms. Data structures such as lists, trees, graphs, and strings, the associated algorithms and their complexity will be treated. Design principles for algorithms such as recursion, divide-and-conquer and dynamic programming will be treated as well.
Knowledge and understanding: Students are able to give examples of data structures and explain do they support program design. Students are able to name what types of standard data structures exist and illustrate their properties. Students are able to describe some standard algorithms and highlight their properties. Students are able to illustrate how to develop and analyse new algorithms.

Applying knowledge and understanding: Students are able to select the appropriate data structure for a given problem. Students are able to propose an algorithm for solving a given problem.

Making judgements: Students are able to justify if and determine how data structures should be applied. Furthermore, students are able to assess whether algorithms are appropriate and efficient.

Communication: Students are able to explain how data structures and algorithms are to be included in program designs.

Learning skills: Students are able to reflect on which data structures and/or algorithms are applicable for each problem.


Exam: ‘Closed Book’ written exam, during the course the students will receive a number of assignments, which can earn them up to a total of one bonus point.

ECTS: 4

Calculus (Code: KEN1440)

Examiner: Dr. A. Briassouli and Dr. O. D’Huys.

Prerequisites: None.

Description: The following subjects will be discussed in Calculus: limits and continuity, differential calculus, inverse and transcendental functions, mean value theorem, integral calculus, sequences and series, introduction to differential equations, introduction to multivariable calculus. In addition to the main facts and concepts, problem-solving strategies will be discussed. Both the intuition behind the concepts and their rigorous definitions will be presented along with simple examples of formal mathematical proofs.

Knowledge and understanding: Student can define, write and explain key facts and concepts involving limits and continuity, can interpret and solve differential calculus, inverse and transcendental functions, mean value theorem, integral calculus, sequences and series, first-order linear differential equations, basics of multivariable calculus.

Applying knowledge and understanding: Students are able to provide examples of instances of problems of specific properties. Students are able to solve problems involving concepts learned in the course, using standard problem-solving strategies. Students are able to understand simple mathematical proofs.

Making judgements: Students are able to analyse a simple problem within the course content and justify the solution methodology they choose. They can summarize this methodology mathematically.

Communication: Students are able to explain their solution strategy in written form and defend their solution strategy in discussion with others.

Learning skills: After successful completion of the course the students will be able both to solve standard problems (constructing graphs of functions, finding extrema of functions, computing limits, summing infinite series etc.) and to apply their knowledge in solving and analysing more complex problems (e.g. in analysis of numerical algorithms).

Study material: Calculus, a complete course, any edition, by R.A. Adams, Addison Wesley Longman and materials provided during the lecture.

Exam: Midterm and final written exams.

ECTS: 4
**ICT and Knowledge Management (Code: KEN1430)**

**Examiner:** Dr. ir. P. Bollen (SBE)

**Tutor:** Dr. ir. P. Bollen (SBE)

**Description:** Knowledge is a fundamental prerequisite in the ability of a person to execute a task. This ability consists of explicit knowledge or information, implicit knowledge or experiences, skills and attitudes.

In this course, we will focus on the (fact-based) conceptual modelling approach in which we also clearly make a link to a relevant contemporary domain in computer science: block chain. We will use the following definition of Knowledge Management:

Knowledge management is an integral approach for the identification, the structuring, the sharing and evaluation of knowledge in the organization.

**Knowledge and understanding:** Students will be able to explain fundamental building blocks of a domain knowledge model. Students can describe the development of the semantic-conceptual modelling approach to knowledge management from the 1970s until now. Students can describe and outline the semantic-conceptual knowledge modelling process.

**Applying knowledge and understanding:** Students will be able to apply the steps in the semantic-conceptual knowledge modelling process on scaled-down examples. Students will be able to apply all steps in the semantic-conceptual knowledge modelling process on ‘real-life’ modelling domains.

**Making judgement:** Students will be able to assess a knowledge domain and determine whether semantic-conceptual modelling can be applied and what domain knowledge and domain examples will be needed to derive the fact types and integrity rules.

**Communication:** Students will be able to write a professional report in which they communicate their findings to a domain problem that they have solved by understanding, assessing and applying the semantic conceptual modelling knowledge they have acquired in (the first part of) this course.

**Learning skills:** Students will be able to apply a professional way of working according to a well-documented methodology.

**Study material:** E-reader

**Exam:** hand-in exercises, mid-term open book exam, final group project

**ECTS:** 4

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**Logic (Code KEN1530)**

**Examiner:** Dr. ir. N. Roos and Dr. O. D’Huys

**Prerequisites:** None. The course appears as a prerequisite for the course Logic for AI.

**Description:** This course deals with three logical systems, namely propositional logic, first-order predicate logic and epistemic logic. The course covers notation systems, syntax and semantics, valid consequences, deduction, semantic tableaux, and proof systems.

**Knowledge and understanding:** Students need to get accustomed to the fundamental concepts of mathematical logical systems (propositional logic and predicate logic) to able to describe information in a logical framework and to reason and prove correctly. Students will get accustomed to the basic concepts of some advanced logical systems (epistemic logic and dynamic logic).

**Applying knowledge and understanding:** Student will apply the reasoning and proof methods learned to small-scale problems and some more complex situations.

**Making judgements:** Students will learn to judge how to reason correctly using mathematical proofs and how to judge which logical system is suitable to solve the problem at hand.
Communication: The chosen syntax of the logical language used must be easily understandable by peers and others experts the logical proofs given must be correct, concise and easily understandable

Learning skills: having learned basic logical concepts and reasoning techniques the students are able to apply them to larger-scale problems

Study material:
- Johan van Benthem, Hans van Ditmarsch, Jan van Eijck, Jan Jaspars, Logic in Action. Edition of February 2014 or later. This is a freely available e-book. Check your Student Portal for the link.

Exam: Written exam; during the course the students will receive three assignments, that, if they receive a sufficient grade, may earn them up to a total of one bonus point.

ECTS: 4

Numerical Mathematics (Code: KEN1540)

Examiner: Dr. P. Collins and Dr. K. Stankova.
Tutor(s): Dr. P. Collins.

Desired Prior Knowledge: calculus, linear algebra

Description: Numerical mathematics is the art of solving mathematical problems with the aid of a digital computer. In this course, we will cover the fundamental concepts of numerical mathematics, including the floating-point representation of real numbers, truncation and round off errors, iterative methods and convergence. We will study the simplest and most important algorithms for core problems of numerical mathematics, namely the solution of algebraic equations, interpolating data by polynomials and splines, numerically estimating derivatives and integrals, solving differential equations, approximating functions by polynomials and Fourier series, solving systems of linear algebraic equations and computing eigenvalues. There will be a strong practical component, with students being expected to write their own numerical code and test the performance and suitability of different methods on various problems.

Knowledge and understanding: By the end of this course, students will have knowledge of the fundamental problems of numerical mathematics and basic techniques for their solution. You will understand issues of efficiency and numerical accuracy, will be able to analyse which numerical methods are likely to perform best on different types of problem, and evaluate whether the results of a given computation are trustworthy. You will be able to write your own code (in MATLAB) implementing basic numerical algorithms. Advanced students will have the skills necessary to adapt existing numerical algorithms and develop new algorithms.

Applying knowledge and understanding: Students will be expected to implement the algorithms covered in the lectures, and apply these to practical problems.

Making judgements: Students will learn how to analyse which numerical methods are likely to perform best on different types of problem, and to evaluate whether the results of a given computation are trustworthy.

Communication: Students will learn the terminology required to discuss numerical algorithms and the results of numerical computations with mathematicians, (social) scientists and engineers.

Learning skills: Students will learn to design, analyse, implement and apply numerical methods.

Study material: Slides, exercise sheets.

Recommended literature: Faires & Burden, “Numerical Methods”.

Exam: Written examination with formula sheet (80%). Computer-based homework exercises (20%+10% Bonus).

ECTS: 4
Software Engineering (Code: KEN1520)
Examiner: Dr. C. Seiler, T. Pepels, M.Sc.
Desired Prior Knowledge: Introduction to Computer Science 1 and 2, Data Structures and Algorithms.
Prerequisites: None
Description: This course introduces students to software design and project management concepts. Students are introduced to multiple techniques they require to work on medium and large-scale projects in professional business and research environments. Students learn how to produce professional, reliable, and cost-efficient software that can be developed in a team, reused, maintained, further evolved, and that is tested professionally. Covered concepts include requirement engineering, project planning, risk management, software evaluation and testing, software engineering processes, design principles, software architectures, design patterns, code review, version control, specifications, debugging, and abstract data types.
Knowledge and understanding: Students learn how to efficiently design and write professional software that meets specifications made by themselves or by customers. Students learn what the essential elements of the software engineering process are. These elements include requirement analysis, design methodologies, implementation strategies, and validation techniques.
Applying knowledge and understanding: Students acquire the skills
• to critically analyse software requirements, software designs, software implementations, and software evaluations.
• to efficiently plan, execute, and monitor progress in-group projects.
• to cooperate better in a group and to participate more effectively as a professional in academia or business environment.
Software engineering is a core activity of knowledge engineers and data scientists. In their professional career DKE students often will act as active programmers, software designers, and project managers where they need to create software as part of a team or act as team leader. The project management skills and tools being taught in this course that help students creating professional and cost-effective software are thus crucial for their career and lay the foundation for further studies of in this field. The knowledge obtained will directly help students for the subsequent semester projects and Bachelor thesis.
Making judgements: Students learn to judge the viability of selected software development methodologies and new developments in design concepts during their career. Students learn to compare design choices and judge their consequences.
Communication: Students learn to discuss and document software developments professionally. Knowledge about widely spread standard software development techniques and about standard design patterns are essential for efficient communication between software developers. Standards in software engineering facilitate cost-effective communication and help to avoid misunderstandings between customers and suppliers, between team leaders and team members as well as between team members.
Learning skills: Students learn to successfully reflect on their project management skills, on how they contribute to a software project as part of a team, and on how to adjust their software engineering approaches to different professional scenarios. Students learn to reflect on and to verify own and others software designs and implementations in a professional manner.
Study material: Lecture material provided during the lecture.
Recommended literature:
• Gamma et al., Design Patterns: Elements of Reusable Object-Oriented Software (1994)

Exam: Written “closed-book” exam at the end of the course. During the course, students receive several graded assignments that can earn them a maximum bonus grade of 1.0.

ECTS: 4

Period 1.6

Project 1–2 (Code: KEN1600)

Examiners: Dr. P. Collins, Dr. ir. N. Roos, Dr. C. Seiler and Dr. Stankova

Coordinator: Dr. K. Schüller

Tutors: Dr. K. Schüller and Dr. O. D’Huys

Prerequisites: In order to participate in this project the student has to have passed two out of four courses from the set: Discrete Mathematics, Linear Algebra, Computer Science I and Computer Science II.

Description: Students work on a project assignment in small groups of about six students. The group composition is announced at the beginning of period 1.4 and remains the same for the whole project. The students are guided through the project by a fixed tutor. The project assignment is divided into three subtasks (one per period) and is strongly related to the content of the courses from period 1.4 and 1.5. In period 1.4, after receiving the assignment for the whole project at the end of week 5, the students work full-time on the project in week 6. In this week, each group meets the tutor twice. In period 1.5, the students continue working on the project, while also having to attend the courses of that period. They meet their tutor approximately once a week. In period 1.6, the students work three weeks full-time on the project and meet their tutor twice a week.

At the beginning of period 1.5 and 1.6, the students have to hand in a planning for the current phase. At the end of each period, the students give a presentation and the source code, presentation and an overview of who did what need to be uploaded to Student Portal. While the presentations at the end of period 1.4 and 1.5 are in front of the examiners and the tutors, the presentations at the end of period 1.6 will additionally be in front of the fellow students. In period 1.6, they furthermore have to hand in a report and attend a product and report examination.

Knowledge and understanding: Interpret the meaning of mathematical models of real-world processes. Gain insight into practical use of software design and development principles. Recognise and relate user-computer interactions to concepts from graphics and user-interface frameworks.

Strengthen knowledge of basic algorithms and methods for specific problems in artificial intelligence and applied mathematics.

Applying knowledge and understanding: Students will be able to design an answer strategy for scientific questions using analytical thinking and logical reasoning and to translate mathematical models to software code. Furthermore, students will be able to implement software to solve problems in applied mathematics by applying numerical methods and artificial intelligence algorithms, formulate computational experiments, and analyse and interpret the results, apply design and development principles in the construction of software systems and use existing software application frameworks for graphics and user interfaces.

Even more so, students will learn to use tools for software project management such as version control systems and issue trackers, identify project goals, deliverables, and constraints. Lastly they will learn how to plan and chair meetings, create notes for minutes, work in a team such that the workload is balanced and plan teamwork by
setting deadlines and distributing tasks.

**Making judgements:** Students will learn to evaluate different mathematical and computational models with respect to their suitability, efficiency and correctness for a specific task.

**Communication:** Students will be able to give a clear and well-constructed presentation, including a demonstration of the product, and with appropriate use of illustrations and/or videos, to offer and respond to questions on and constructive criticism of presentations. Furthermore, they will learn to write a project report according to the structure of an academic article, submit arguments in exact sciences, with appropriate use of formulae and figures. They learn to cite published sources in the project report according to the academic guidelines. Additionally, students will learn to structurally inform stakeholders on project progress and effectively communicate with project group members about task division, planning and project deadlines, effectively communicate with group members by listening to others’ ideas; be contactable include others in the discussion. It will be important to cooperate in a group to reach a consensus view, communicate in the English language, elicit and evaluate relevant scientific background information.

**Learning skills:** Reflect on one’s own academic abilities and functioning in a team.

**Study material:** Project manual project 1-2, Maastricht University, DKE.

**Assessment:** Final grade = 0.15 x grade of phase 1 + 0.15 x grade of phase 2 + 0.7 x grade of phase 3, where grade of phase 3 = 40% grade for product + 40% grade for report + 10% grade presentation + 10% grade project management. Missing project meetings and skill classes will cause an individual grade reduction of the final grade. Furthermore, the examiners may deviate from the group grade for individual students if they either contributed significantly more or less than other group members.

**Skill classes:**

**Team Dynamics 3 (period 1.4)**
This team dynamics workshop aims to provide you with a deeper awareness, insight and practice in effective team collaboration & co-creation. During this introduction workshop, you and your project team will draw up a team charter (contract) to initiate effective group collaboration in project 1-2.

**Referencing Tools (period 1.4)**
During this class, you will learn how to 1) use referencing tool EndNote, 2) to collect and keep track of your references, and 3) to easily and correctly insert them into your project report documents as in-text citations, footnotes or a bibliography in any of a large number of citation styles.

**Academic Writing (period 1.5)**
In the project skills meetings you will explore the key structure of your report, as well as key points of Academic Writing at Maastricht University. Areas of focus include structure of paper, linguistic aspects of writing in English, presenting information logically and citation and reference procedures.

**Advanced Presenting (period 1.5)**
This second period will focus on presentation skills and techniques. This introduction helps with public speaking and prepares you for project report presentations. Areas of focus include structure of a presentation, public speaking techniques and enunciating, language aspects to remember while planning a presentation, and the dos and don’ts expected by Maastricht University.

**Team Dynamics 4 (period 1.5)**
In this skill class, you are going to evaluate the team collaboration and communication during project 1-2 by means of interactive exercises.

**ECTS:** 6
2.2 Curriculum of 2nd Year of the Bachelor’s Programme

Year 2

<table>
<thead>
<tr>
<th>Period</th>
<th>Courses</th>
<th>ECTS</th>
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<tbody>
<tr>
<td>2.1</td>
<td>Databases (KEN2110)</td>
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<td></td>
<td>Probability and Statistics (KEN2130)</td>
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<td></td>
<td>Graph Theory (KEN2220)</td>
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<td></td>
<td>Project 2-1 (*)</td>
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<tr>
<td>2.2</td>
<td>Reasoning Techniques (KEN2230)</td>
<td>4</td>
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<td>Machine Learning (KEN2240)</td>
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<td>Linear Programming (KEN2520)</td>
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<td></td>
<td>Project 2-1 (*)</td>
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<td>2.3</td>
<td>Project 2-1 (KEN2300)</td>
<td>6</td>
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<tr>
<td>2.4</td>
<td>Mathematical Modelling (KEN2430)</td>
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<td>Human Computer Interaction and Affective Computing (KEN2410)</td>
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<td></td>
<td>Theoretical Computer Science (KEN2420)</td>
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<td></td>
<td>Project 2-2 (*)</td>
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<tr>
<td>2.5</td>
<td>Philosophy &amp; Artificial Intelligence (KEN2120)</td>
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<td></td>
<td>Simulation and Statistical Analysis (KEN2530)</td>
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<td></td>
<td>Natural Language Processing (KEN2570) (Elective**)</td>
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<td></td>
<td>Introduction to Image and Video Processing (KEN3238)(Elective**)</td>
<td>4</td>
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<tr>
<td></td>
<td>Project 2-2 (*)</td>
<td>-</td>
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<tr>
<td>2.6</td>
<td>Project 2-2 (KEN2600)</td>
<td>6</td>
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</table>

(*) Project 2-1 will start in period 2.1 with weekly meetings; Project 2-2 will start in period 2.4 with weekly meetings. The credits for the projects will become available at the end of period 2.3 and 2.6 respectively.  

** Electives: In case students have passed both electives of period 2.5, either the course Natural Language Processing or Introduction to Image and Video Processing can replace 1 of the third year electives

For each period, we will give a short explanation of the various parts. Before the start of each period, the students will receive detailed information about the content, the study material, the teaching form, the schedule, and the examination method.

Period 2.1

Databases (Code: KEN2110)  
Examiner: Dr. D. Rafailidis  
Desired Prior Knowledge: Introduction to Computer Science 1 and 2.  
Prerequisites: None.  
Description: This course will cover data modelling, the concepts and theory of the relational data model, and the widely used programming language SQL. These concepts will be applied in case studies.  
Knowledge and understanding: Students will be able to describe the basic concepts of databases, explain the fundamental concepts of database management systems, and query languages.  
Applying knowledge and understanding: Students will be able to explain the proper database design based on the users’ requirements, indicate possibilities and limitations of each database type. In addition, students will be able to combine fundamental Computer Science architectures and cooperate in a group to design and
construct a database.

Making judgements: Student will be able to analyse and justify a practical database problem, examine different design models, and refine database models based on real-life use cases.

Communication: Students will be able to summarize the basic entities and relationships involved in the database design, and communicate with the end users to fetch the database requirements.

Learning skills: Students will be able to reflect on certain solutions of the databases design and implementation, assess the correctness of the database models and identify follow-up literature, beyond the teaching material of the course.


Exam: Written exam (75%) + practical assignment (25%)

ECTS: 4

**Probability & Statistics (Code: KEN2130)**

Examiner: Dr. C. Seiler

Desired Prior Knowledge: Discrete Mathematics and Calculus

Prerequisites: None.

Description: Probability theory is the mathematical branch that focuses on experiments whose outcomes are determined by chance. Statistics is about collecting and analyzing numerical data in order to infer properties of the whole population. In this course, we will discuss the following topics:

- Probabilities, conditional probabilities, random variables, discrete and continuous probability distributions, joint probability distributions.
- Expectation, variance, standard deviation, covariance and independence.
- Several discrete and continuous probability distributions.
- Random samples, unbiased estimators, the central limit theorem
- Verification of hypotheses and confidence intervals

Knowledge and understanding: In this course, the students obtain tools to define random variables and identify probability distributions in a wide range of probabilistic experiments. Furthermore, they know which procedure is most appropriate to find an answer to a given statistical question.

Applying knowledge and understanding: Students are capable of calculating probabilities, expectations, variances and related quantities in a wide range of probabilistic experiments. Furthermore, they can estimate statistical quantities and perform statistical tests to extract information from data sets.

Making judgements: Students are able to critically analyse probabilistic experiments and statistical inferences and are able to decide whether to accept or reject statistical hypotheses.

Communication: The students will be able to communicate their conclusions and the underlying rationale to expert and non-expert audiences.

Learning skills: Students are able to use elements from probability theory and statistics in other domains in order to increase one’s knowledge.

Study material: Hwang and Blitzstein, Introduction to Probability (2019, second edition)

Exam: Written exam

ECTS: 4
Graph Theory (Code: KEN2220)
Examiner: Dr. M. Mihalák
Desired Prior Knowledge: Discrete Mathematics; Data Structures and Algorithms
Prerequisites: None
Description: A graph is simply a collection of points, some of which are joined by lines. This deceptively simple structure is one of the cornerstones of both theoretical and applied computer science. A great many problems that arise in the real world can be modeled as graph problems. Several classical examples include the problem of finding the shortest route between two cities, of maximizing flow in a network of pipelines, or of finding an optimal pairing between producers and consumers. In this course we will look at both the algorithmic/applied side of graph theory and its more abstract mathematical foundations, because the latter is often important for understanding the former. We will cover topics such as paths, tours, trees, matchings, flows and colorings.
Knowledge and understanding: Students will have a solid overview of the basic concepts and results of (applied) graph theory, including the main mathematical tools to argue about graphs. Students will have the tools to model and analyze various real-world problems using graphs.
Applying knowledge and understanding: Students will be able to recognize when a problem can be modeled with graphs, and whether the problem can be efficiently solved using standard or slightly adjusted graph-theoretic algorithms.
Making judgements: Students will be able to formulate a given (sub)problem as a graph-theoretic problem, argue why the formulation is correct, and they will be able to judge the feasibility of existing algorithmic solutions.
Communication: Students will be able to explain, in the language of graph theory, how a problem at hand can be modelled and solved.
Learning skills: Students will enhance their study skills such as time management, effective reading, critical thinking and reading, exact and unambiguous writing and formulation of ideas and statements, and reflection on marked (graded) work. Along the way, students will improve general learning skills such as self-motivation, careful listening and giving instructions, and openness to new knowledge.
Study material: Appropriate material will be provided during the course.
Recommended literature: None.
Exam: Written exam (80% of the final grade). Weekly graded exercises (20% of the final grade).
ECTS: 4

Period 2.2
Machine Learning (Code: KEN2240)
Examiner: Dr. E.N. Smirnov, Dr. E. Hortal Quesada and Dr. M. Popa
Desired Prior Knowledge: Introduction to Computer Science 1, Calculus, Linear Algebra, Logic, Probability and Statistics
Prerequisites: None
Description: Machine learning is a major frontier field of artificial intelligence. It deals with developing computer systems that autonomously analyse data and automatically improve their performance with experience. This course presents basic and state-of-the-art techniques of machine learning. Presented techniques for automatic data classification, data clustering, data prediction, and learning include Decision Trees, Bayesian Learning, Linear and Logistic Regression, Recommender Systems, Artificial Neural Networks, Support Vector Machines, Instance-based Learning, Rule Induction,
Clustering, and Reinforcement Learning. Lectures and practical assignments emphasize the practical use of the presented techniques and prepare students for developing real-world machine-learning applications.

**Knowledge and understanding:** After successful completion of the course, students will be able to describe and explain the basic machine learning algorithms. Students will understand the mathematical foundation of machine learning algorithms and how mathematical methods are successfully combined to obtain the variety of machine learning algorithms that are currently available.

**Applying knowledge and understanding:** Students will acquire the knowledge able to apply, formulate, and validate techniques from machine learning and to apply basic machine learning algorithms on real-life problems. Students will be able to implement machine-learning algorithms in software and to apply existing machine learning software implementation to datasets. Students will have the necessary knowledge to design, implement, and apply data processing systems that autonomously extract information from data, interpret results, and make decisions.

**Making judgements:** Student learn how to critically analyse real-world problems, to select appropriate machine learning techniques according to the specific problem, and to predict the consequences of their choices. After successful completion of the course, students gain the ability to judge which problems can be solved better and to which extend through the application of machine learning techniques. Students obtain an awareness of and responsibility for ethical and social consequences of developments in and application of machine learning.

**Communication:** The skills acquired during the course will allow students to present the results of different stages of the application of machine-learning techniques to specialists or non-specialists.

**Learning skills:** After successful completion of the course, students can analyse, adapt, design, implement, and critically reflect on machine-learning algorithms and tools. Students also obtain the critical fundamental skills and knowledge to study further advanced machine learning techniques in the professional literature.

**Study material:** Lecture material provided during the lecture.

**Recommended literature:**
- H. Blockeel, Machine Learning and Inductive Inference (course text), Uitgeverij ACCO, 2012.

**Exam:** Written “open-book” exam at the end of the course. During the course, students receive several graded assignments that can earn them a maximum bonus grade of 1.0.

**ECTS:** 4

**Linear Programming (Code: KEN2520)**

**Examiner:** Dr. S. Kelk.

**Desired Prior Knowledge:** Linear Algebra.

**Prerequisites:** None

**Description:** A linear program is very different to, say, a Java program. It simply consists of a linear objective function (of potentially very many variables) and a set of linear inequalities. The goal is to find values of the variables, which maximize or minimize the objective function, subject to all the inequalities being satisfied. Linear programs - even very large linear programs - can be solved extremely quickly, in both theory and practice. The model is also expressive enough to capture a large number of real-world
problems. These two factors explain the fundamental role of linear programming in operations research, computer science, economics, management and many other fields. The course consists of an in-depth study of the simplex algorithm (a standard algorithm for solving linear programs), duality theory, and sensitivity analysis. Examples from practice illustrate the power of the model and teach the student the skill of modelling. Practical aspects of linear programming (e.g. use of software packages for solving linear programs, and integration with languages such as Java) are also considered.

**Knowledge and understanding:** Students will be able to identify which real-world optimization problems can be formulated as linear programs. Students will be able to describe the mathematical foundations of the Simplex algorithm for solving linear programming, and articulate how these foundations impact upon the performance of the Simplex method in practice. Students will recognize the power and importance of duality theory for reasoning about the behaviour of linear programs (in particular with regard to sensitivity analysis). Students will be able to exhibit an awareness of non-Simplex paradigms for solving linear programs (interior-point methods) and be able to recount the importance of the linear programming model in operations research and applied mathematics.

**Applying knowledge and understanding:** Students will be able to 
1) translate mathematical models into linear programs, 
2) to apply the Simplex method by hand to solve small linear programs, 
3) to show how the Simplex method behaves in normal and exceptional cases, 
4) to manipulate the algebra underpinning the Simplex method, 
5) to combine insights from this algebra and primal-dual relations to make rigorous statements about the (sub)optimality of solutions to linear programs, 
6) to argue how small changes to linear programs impact upon their optima (sensitivity analysis), 
7) to explain key differences between the Simplex method and interior-point methods, and to 
8) leverage linear-to program arguments when developing simple algorithms for combinatorial optimization problems.

**Making judgements:** Students will be able to distinguish between mathematical models that can and cannot be cast as a linear program. Students will be able to contrast and compare the behaviour of the Simplex algorithm with interior-point methods. Students will be able to select, out of a large range of algebraic and duality-based instruments, appropriate tools for making rigorous statements about linear programs.

**Communication:** Students will be able to formulate linear programs and defend their correctness. Students will be able to clearly articulate and defend algebraic and duality-based arguments concerning linear programs.

**Learning skills:** By the end of the course, students will be able to autonomously and critically reflect upon the appropriateness of the linear programming paradigm for tackling optimization problems arising in practice and be able to assess the correctness of mathematical arguments pertaining to linear programming. Students will be able to identify follow-up literature, which goes beyond the scope of the material presented in the course.


**Recommended literature:** students are beforehand encouraged to refresh their knowledge of: (unique) solutions of systems of linear equations, matrix inversion, and matrix rank.

**Exam:** Written exam and optional weekly bonus exercises (the results of which are added to your exam score, up to 10%).

**ECTS:** 4
**Reasoning Techniques** (Code: KEN2230)

**Examiner:** Dr. C. Sironi and Prof. dr. M.H.M. Winands

**Desired Prior Knowledge:** Introduction to Data Science and Artificial Intelligence; Logic.

**Description:** Central in this course is how, based on available data, new knowledge and information can be obtained using reasoning processes. The following four techniques are discussed:

1. **Problem solving using search:** problem types, blind-search methods, informed-search methods, comparison of search methods, games as search problems, minimax, alpha-beta pruning, chance games.

2. **Reasoning using logic:** syntax, semantics, and inference in propositional logic and in first-order logic, situation calculus, forward and backward reasoning, completeness, resolution.

3. **Planning:** planning in situation calculus, representation of states, goals and operators, state space and plan space, partially ordered planning.

4. **Reasoning with uncertainty:** uncertainty and probability theory, conditional probability, the Rule of Bayes, semantics of and inference in belief networks, the principle of maximum utility, multi-attribute utility, influence diagrams.

**Knowledge and understanding:** Students learn to understand how problems can be represented as search problems, as logical problems, as planning problems or as problems involving uncertainty and get accustomed to reasoning methods to solve problems of all four types mentioned above.

**Applying knowledge and understanding:** Students learn to apply the reasoning methods learned to toy problems and some more complex situations.

**Making judgements:** Students learn to judge which type of knowledge representation is suitable for the problem at hand and which reasoning technique is suitable to solve the problem at hand.

**Communication:** the knowledge representation used and reasoning technique chosen must be easily understandable by peers and others experts.

**Learning skills:** Students are able to critically reflect on their own and other’s chosen representations and used reasoning methods.


**Exam:** Written exam; during the course the students will receive three assignments, that, if they receive a sufficient grade, may earn them up to a total of one bonus point.

### Period 2.3

**Project 2–1** (Code: KEN2300)

**Examiners:** Dr. J.W.H.M. Uiterwijk and Dr. M. Mihalák.

**Coordinator:** Dr. K. Schüller

**Tutors:** Dr. K. Schüller and Dr. M. Musegaas

**Prerequisites:** Students must have passed Project 1–1. Furthermore, the student has to have passed at least two out of the following three courses: Introduction to Computer Science 1, Introduction to Computer Science 2, and Data Structures and Algorithms. This project is a prerequisite for Project 3–1.

**Description:** Students work on a project assignment in small groups of about six students. The group composition is announced at the beginning of period 2.1 and remains the same for the whole project. The students are guided through the project
by a fixed tutor. The project assignment is divided into three subtasks (one per period) and is related to the content of the courses from period 2.1 and 2.2. In periods 2.1 and 2.2, the students work on the project, while also having to attend the courses of these periods. They meet their tutor approximately once a week. In period 2.3, the students work three weeks full-time on the project and meet their tutor twice a week. At the beginning of each period, the students have to hand in a planning for the current phase. At the end of each period, the students have to give a presentation and the source code, presentation and an overview of who did what need to be uploaded to Student Portal. While the presentations at the end of period 2.1 and 2.2 are in front of the examiners and the tutors, the presentations at the end of period 2.3 will additionally be in front of the fellow students. In period 2.3, they furthermore have to hand in a report and attend a product and report examination.

**Applying knowledge and understanding:** Students will learn to concretize project assignment and construct and maintain a planning. Furthermore, they will learn formulating, selecting and validating models for the game chosen and collect and interpret experimental data with evaluation metrics. Lastly they will improve their ability to plan and chair meetings, create notes for minutes, work in a team such that the workload is balanced and plan teamwork by setting deadlines and distributing tasks.

**Making judgement:** After completing this project, students will be able to compare and criticize results, position them in terms of the literature diagnose limitations and formulate a discussion

**Communication:** Students will be able to write a scientific paper that describes the project, explains the methods, summarizes the outcomes, discusses them and makes the conclusions. Students will be able to present and defend project in English and coordinate project progress in project meetings

**Learning skills:** Students will be able to reflect on the progress of the project and study relevant literature to solve problem at hand

**Study material:** Project manual project 2-2, Maastricht University, DKE.

**Assessment:** Final grade = 0.15 x grade of phase 1 + 0.15 x grade of phase 2 + 0.7 x grade of phase 3, where grade of phase 3 = 45% grade for product + 35% grade for report + 10% grade presentation + 10% grade project management. Missing project meetings and skill classes will cause an individual grade reduction of the final grade. Furthermore, the examiners may deviate from the group grade for individual students if they either contributed significantly more or less than other group members.

**Skill Classes:**

- **LaTeX (period 2.1)**
  The students will learn to set up a LaTeX environment on their computers. Furthermore, they will learn the basic aspects of LaTeX such as figures, tables, referencing and formulas. In the end of the course, the students will be able to write reports and articles with LaTeX.

- **Systematic Search Strategies (period 2.2)**
  During this class, you learn more about how to elicit and evaluate relevant scientific background information for writing a project report. By the end of this class, you will be able to apply systematic search strategies - usable in multiple databases.

- **Legal aspects of Data Science (period 2.2)**
  In this session, you will get an introduction into the basic principles of the GDPR. By the end of this session, you will be able to grasp the key actors, concepts and obligations of the GDPR, and develop awareness and understanding of the legal requirements you will encounter in your professional career.

**ECTS:** 6
Period 2.4

Theoretical Computer Science (Code: KEN2420)

Examiner: Dr. G. Stamoulis
Desired Prior Knowledge: Introduction to Data Science and Artificial Intelligence, Discrete Mathematics, Data Structures and Algorithms.
Description: This course explores the theoretical underpinnings of computing by investigating algorithms and programs casted as language recognition problems. The influence of the theory on modern hardware and software system design is demonstrated. The following subjects will be treated: mathematical foundations, alphabets and languages, finite automata and regular languages, Turing machines, acceptance and decidability, recursive functions and grammars, time complexity classes, NP problems, NP-completeness
Knowledge and understanding: Students will learn to comprehend the inherent complexity of problems and be able to motivate why some problems are inherently more difficult than others are. They will learn to have insight into how complex problems can be solved efficiently and will be able to classify such problems into a language hierarchy and complexity classes. Furthermore, students will be able to apply the tools needed for such classification
Applying knowledge and understanding: students will be able to apply the theory learned to solve small-scale problems
Making judgements: Students will learn to judge which problems are decidable and efficiently solvable and to judge which technique is suitable to solve the problem at hand.
Communication: The knowledge representation used and technique from complexity theory chosen must be easily understandable by peers and others experts
Learning skills: The student will learn to reflect on own one’s and other’s thoughts on complexity and solvability of problems.
Exam: Written exam; during the course the students will receive three assignments, that, if they receive a sufficient grade, may earn them up to a total of one bonus point.
ECTS: 4

Mathematical Modelling (Code: KEN2430)

Examiner: Dr. J. Karel and Prof. dr. ir. R. Peeters.
Desired Prior Knowledge: Linear Algebra, Calculus, Matlab.
Prerequisite: None.
Description: Mathematical modelling is of great importance for solving practical problems by casting them into a form suitable for the use of mathematical techniques. In this course, a number of basic topics are discussed. First, attention is paid to a framework for mathematical modelling. Then we focus on some widely used model classes from engineering, in particular on the class of linear time-invariant dynamical models. These are described by linear difference equations (in discrete time) or linear differential equations (in continuous time). Alternative model descriptions that are discussed are transfer functions (in the frequency domain) obtained with the z-transform and the Laplace transform respectively; and state-space models, which may or may not involve canonical forms. Some further topics receiving attention are the concepts of stability, sinusoidal fidelity, Bode diagrams, the interconnection of subsystems, and the technique of pole placement by means of state feedback. The subject matter is clarified through exercises and examples involving practical applications. Also, relevant functionality in Matlab is introduced, which offers a powerful instrument for analysing linear dynamic models.
Knowledge and understanding: Being able to formulate linear dynamical models, state properties and define representations. Identify frequency domain properties of systems and relate them to applications in signal processing.

Applying knowledge and understanding: Being able to construct elementary mathematical models. Perform model analysis and extract model properties. Employ various model representations and choose the most appropriate one. Compute state-feedback control.

Making judgements: To recognize what are the important aspects to consider when building a mathematical model. Decide on stability of models.

Communication: Being able to convey properties of models to specialists and non-specialists.

Learning skills: Being able to independently find Matlab functionality to solve basic problems in systems theory.

Study material: Lecture notes.


Exam: Written exam and assignments and/or bonus assignments.

ECTS: 4

Human Computer Interaction and Affective Computing (Code: KEN2410)

Examiner: Dr. S. Asteriadis.

Desired Prior Knowledge: Machine Learning, Probabilities and Statistics.

Description: Human -Computer Interaction (HCI) is the study of interaction between people (users) and computers. It is often regarded as the intersection of computer science, behavioural sciences, design and several other fields of study. Interaction between users and computers occurs at the user interface, which includes both software and hardware; for example, characters or objects displayed on a personal computer’s monitor and input received from users via hardware peripherals such as keyboard, mouse and web cameras. This course also covers Affective Computing, a new branch of HCI that places emphasis on user emotions and personality. Affective Computing attempts to bring emotions into intelligent interfaces that interact with humans and see how they can have a positive and constructive impact in human-machine interactions.

Knowledge and understanding: The course shows guidelines for the design, implementation and evaluation of HCI systems and addresses interaction styles with the user. Students will be able to identify ways to involve user emotion, personality and cognition in the design of an HCI intelligent interface.

Applying knowledge and understanding: Students will be able to apply stages of successful user interface design. Formulate user interview techniques, rapid prototyping, design interfaces and conclude on user needs in business-like scenarios.

Making judgements: Students will show awareness of how to involve the user in the design procedure and solve HCI problems based on judgement analysis of user evaluations and interviews.

Communication: Students will be able to present their prototypes, communicate and defend their results. They will be able to examine existing interfaces and they will ground their findings based on HCI guidelines.

Learning skills: Students will be able to describe a course of action for designing human-centric systems, applicable in a variety of social and business practices. They will be able to establish links among Machine Intelligence, Affective Interactions and Statistical analyses in the design of user-centered interfaces. They will be able to use these skills in constructing personalized solutions, quantitatively analyse and apply user needs, design
emotion-capturing techniques by using computational models of affect.

**Study material:** Lecture Slides and other sources that will be made available.

**Recommended literature:**
- Coursera video lectures of Scott Klemmer and accompanying slides.

**Exam:** Individual (90% of final grade) and group (10% of final grade) assignments.

**ECTS:** 4

**Period 2.5**

**Philosophy & Artificial Intelligence** (Code: KEN2120)

**Examiner:** Dr. D.M. Cressman.

**Prerequisites:** none.

**Description:** One of the characteristics of scientific knowledge is the translation of natural phenomena into quantitative or mathematical data – the book of nature, Galileo wrote, is written in the language of mathematics. Over the course of the twentieth and twenty-first century, this desire to understand the world through the logic of mathematics has been extended beyond the natural world to include such things as human consciousness, learning, and intelligence. Indeed, the foundation of what is called ‘artificial intelligence’ is the pursuit of replicating human consciousness and intelligence through mathematical models and formulas. In this course we will examine these issues from a philosophical perspective, beginning with a basic overview of the philosophy of science with an emphasis on quantification and then moving on to study philosophical issues that have developed out of the pursuit of artificial intelligence. We will begin with classic thinkers in the field like Alan Turning, Hubert Dreyfus, and Joseph Weizenbaum and continue through to contemporary philosophical studies of cutting edge attempts to develop types of machine learning that aim to mimic human forms of learning.

**Knowledge and understanding:** At the conclusion of this course, students should be able to demonstrate knowledge of the following topics through written essays:
- The history of computing and artificial intelligence
- The history and philosophy of scientific knowledge with an emphasis on Kuhn’s theories of scientific paradigms
- Historical and philosophical theories of technology and society
- The philosophical presuppositions of artificial intelligence

**Applying knowledge and understanding:** Students will be able to draw upon both lectures and readings to write an essay that exhibits critical reflections on conventional and naive notions of instrumentalism, technological determinism, and functionalism by persuasively arguing for a contextual approach that highlights the contingency and flexibility of design and meaning.

**Making judgements:** Students are asked to select relevant passages from texts that contribute to the argument that they make in the essay. This will be graded. In tutorials, students are asked to make decisions about specific problems (i.e. self-driving cars, Turing tests). This is not graded.

**Communication:** During tutorials, students present their work orally to their classmates.

**Learning skills:** Students will be able to articulate and solve problems in groups.
Students will also be expected to engage with a number of theories concerning computation and artificial intelligence through different texts and will be asked to reflect upon and critique these theories.

**Study material:** Selected texts will be made available.

**Exam:** Take home exam (essay).

**ECTS:** 4

### Simulation and Statistical Analysis (Code: KEN2530)

**Examiner:** Dr. J. Karel and Dr. A. Wilbik

**Prerequisites:** none

**Desired Prior Knowledge:** Probability & Statistics, Calculus, Matlab, and Java.

**Description:** Mathematical simulation is concerned with studying processes and systems. Uncertainty can be an important factor and has to be modelled properly. After modelling a complex system, various scenarios can be simulated, using Monte Carlo simulation, to gain insight. The results need to be properly interpreted and uncertainty has to be reduced. The modelling, implementation, analysis and technical aspects will be discussed as an introduction in this field. Emphasis will be on discrete event simulation and the statistical analysis of the output of simulation studies, where topics are: modelling, Poisson processes, random number generators, selecting and testing input distributions, generating random variates, statistical analysis of experiments and variance reduction. Practical exercises will be used to place the techniques in context.

**Knowledge and understanding:** Define concepts of simulation and discrete event simulation. Explain techniques underlying mathematical simulation. Recall and explain methods for analysing simulation output and efficient simulation including their assumptions, justify why they are important, and match them to simulation design. The use of Knowledge and understanding: Being able to model a system in a structured manner, to design and implement simulators for systems, and to collect data from these simulations. In addition, you will be able to employ techniques underlying mathematical simulation and apply methods for analysing simulation output and efficient simulation.

**Making judgements:** Being able to choose and motivate alternative techniques underlying mathematical simulation. Choose, motivate and contrast methods for analysing simulation output and efficient simulation.

**Communication:** Being able to convey the phases of a specific simulation study to non-experts.

**Learning skills:** The ability to independently learn to handle large-scale simulation. To identify shortcomings in data analysis.

**Study material:** Simulation Modeling and Analysis (5th edition) - Averill Law


**Exam:** Written exam and assignments and/or bonus assignments

**ECTS:** 4

### Natural Language Processing (optional course) (Code: KEN2570)

**Examiner(s):** Dr. J. Niehues

**Tutors(s):** None.

**Desired Prior Knowledge:** Introduction to Computer Science 1 and 2, Probability and Statistics, Machine Learning

**Prerequisites:** None

**Description:** Watson won Jeopardy. Siri can tell me when I need an umbrella. But how do they work? Over the past decade, Natural Language Processing (NLP) was revolutionized...
by statistical, probabilistic and machine learning methods. NLP addresses fundamental questions at the intersection of human language and machine learning. How can computers acquire, understand and produce language? How can computational methods give us insight into observed human language phenomena? How to make sense of the vast amounts of information available online in free, unstructured form? In this course students will learn how computers can learn useful text/language representations and how different tasks (language modelling, text classification, information extraction, sequence labeling, etc.) can be used for solving different complex problems (spelling correction, spam detection, search engine design, opinion analysis, summarization, question-answering, etc.). Open NLP problems (such as evaluation or interactive dialogue systems) and the effect of deep learning on NLP will be discussed.

**Knowledge and understanding:** By the end of the course, students are able to acquire the basic text and language processing aspects. Furthermore, students are able to describe basic NLP problems, tasks and methods.

**Applying knowledge and understanding:** Students are able to demonstrate how to tackle a text/language problem and to formulate, design and implement a NLP system. Students are able to suggest when a problem’s complexity requires an NLP solution.

**Making judgements:** Students are able to pose questions and define problems in different domains (e.g. social sciences) and contexts (e.g. business) that include language/text data. Furthermore, students are able to judge which tools are applicable for solving these problems and to decide a course of action in accordance with ethical and social consequences.

**Communication:** Students are able to outline an approach in real organizational problems, which require NLP and are able to demonstrate, present and communicate a solution to a NLP problem.

**Learning skills:** Students are able to master and choose the appropriate basic programming tools for NLP and are able to follow up on literature that will allow them to build complete NLP models.

**Study material:** Handouts

**Recommended literature:**


2) Chris Manning and Hinrich Schütze, Foundations of Statistical Natural Language Processing, MIT Press. Cambridge, MA. 1999

**Exam:** Practical individual assignments (30%) + Group Project (20%) + Open-Book Written Exam (50%).

**ECTS:** 4

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**Introduction to Image and Video Processing (optional course)**

(Code: KEN3238)

**Examiner:** Dr. Alexia Briassouli

**Desired prior knowledge:** Calculus, Linear Algebra, Machine Learning.

**Prerequisites:** None.

**Description:** Image and video processing is everywhere around us, in smartphones, robotics, medicine, security systems, microscopy, remote sensing, video games, travel, shopping, environmental management and many other applications. Image and video processing is based on principles of signal processing, extended to multiple dimensions. In this class students will have a short introduction to basic 2D signals and systems, sampling, convolution. Color domain processing in different spaces and its relevance to our visual perception system will be presented. We will learn about linear and
non-linear filtering in the spatial and frequency domains (Fourier, DCT), their relation and applications like enhancement, noise estimation and removal, compression, restoration. Compression standards for image and video and their relevance to frequency transformations will be presented. Video analysis will include methods for motion estimation, segmentation and introduction to action recognition, and video standards. Lab examples and homeworks will accompany the classes.

**Knowledge and understanding:** Students will gain an in-depth knowledge of image and video processing methods used all around us by understanding the underlying theoretical foundations and obtaining insights into their role in numerous applications. They will understand the mathematics and algorithms underlying image and video analysis methods and see their results in practice through mini projects.

**Applying knowledge and understanding:** The students will be able to immediately apply basic image and video processing concepts to real world problems in the labs. They will also have the opportunity to further demonstrate their understanding by implementing them in mini projects.

**Making judgements:** By understanding fundamental signal/image processing principles and seeing them work in practice, the students will be able to understand how and where to implement and develop image and video analysis algorithms, as well as build complex systems using and extending them.

**Communication:** Assessment will be based on mini projects with topics on image/video processing chosen by the instructor (the students themselves are also free to suggest subjects relevant to the class). The goal will be for them to learn to carry out independent work, solve realistic problems based on the class material, as well as effectively communicate their motivation and results both to a general and expert audience by providing interesting demos, presentations and clearly structured reports.

**Learning skills:** Students will be able to carry out basic image and video analysis algorithms, understanding and implementing the mathematics behind them. They will solve small research questions within mini projects and will become familiarized with the newer toolboxes and libraries, mostly in – but not restricted to - Matlab and Python.

**Study material:** Lecture slides and provided material. Gonzalez & Richard E. Woods, “Digital Image Processing “.

**Recommended literature:** Computer Vision: a Modern Approach. D. A. Forsyth, J. Ponce (online).

**Exam:** Three assignment (100%).

**ECTS:** 4

**Period 2.6**

**Project 2–2 (Code: KEN2600)**

**Examiners:** Dr. J. Niehues and Dr. A. Briassouli.

**Tutor(s):** Dr. K. Schüller.

**Prerequisites:** Students should have passed Project 1-2. Furthermore, the student has to have passed at least two out of the following three courses: Introduction to Computer Science 1, Introduction to Computer Science 2, and Data Structures and Algorithms. This project is not a prerequisite for another project / course.

**Description:** In small groups, guided by the teachers of the subjects concerned and by the tutor(s), a project is carried out. This project will be strongly related to the content of the subjects from period 2.4 and 2.5. The specific project assignment will be announced in time.
Applying knowledge and understanding: Students will learn to concretize project assignment and construct and maintain a planning. Additionally, they will learn formulating, selecting and validating models for a concrete problem at hand and to collect and interpret data with evaluation metrics.

Making judgement: After completing this course successfully, students will be able to compare and criticize results, position them in terms of the literature, diagnose limitations and formulate a discussion.

Communication: Students will be able to write a scientific paper that: describes the project, explains the methods, summarizes the outcomes, discusses them and makes the conclusions. Furthermore, student will be able to present and defend project in English. Coordinate project progress in project meetings.

Learning skills: Students will learn to reflect on the progress of the project and study relevant literature to solve problem at hand.

Study material: Period book 2.4-2.6, Maastricht University, DKE.
Project skills Period 4: Interviewing and Questionnaires
In these project skills meetings we strategically look at how to formulate questions and questionnaires that could be used in market research. Some language aspects will also be reviewed to help with the question writing and formulation process.

Course Coordinator: Sarah Crielesi, MA
Tutor(s): TBA
Study material: Will be given out in lectures and tutorials.
Assessment: Attendance, activity participation in tutorials.

Project skills Period 5: Project Integration: How to Present Your Project
This second period will review both written and presentation skills and techniques specifically in argumentative and persuasive settings. This review of basic skills and further strengthening of your written and speaking skills help prepare you for your presentation, your written report, and can also be applied to future research funding applications.

Course Coordinator: Sarah Crielesi, MA
Tutor(s): TBA
Study material: Will be given out in lectures and tutorials.
Assessment: Attendance and activity participation in tutorials
Exam: The project will be assessed based on report, product, and presentation.
ECTS: 6

2.3 Curriculum of the Third Year of the Bachelor’s Programme

The first semester of the third year allows you to make your own selection of subjects in the field of artificial intelligence, data science and operations research, the core areas of the study of Data Science & Artificial Intelligence. In period 1 and 2, you can choose 6 out of 12 optional courses. The first semester of year 3 has the same structure in the first and the second year; there are two periods of eight weeks and one period of four weeks. There is also a project in period 3.3. Alternatively, students can choose to study the first semester of the third year at a university abroad (for partner universities see Section 1.4). Please refer to the separate ‘study abroad guide’ and contact your study advisor for more information about study abroad. In periods 3.4, 3.5 and 3.6, you will work on a bachelor’s thesis about a data science and knowledge engineering topic of your interest.

In the diagram below we indicated the courses offered by DKE and the study load expressed in credits (ECTS). A credit stands for about 28 hours of studying time (teaching meetings and private study).
Year 3

Period 3.1*  
- Prolog (KEN3234)  4
- Semantic Web (KEN3140)  4
- Game Theory (KEN3130)  4
- Robotics and Embedded Systems (KEN3236)  4
- Computer Security (KEN2560)  4
- Digital Society (KEN3111)  4
- Project 3-1 (**)  

Period 3.2*  
- Large Scale IT and Cloud Computing (KEN3239)  4
- Logic for Artificial Intelligence (KEN3231)  4
- Parallel Programming (KEN3235)  4
- Introduction to Bio-Informatics (KEN3440)  4
- Software and Systems Verification (KEN3150)  4
- Introduction to Quantum Computing (KEN3241)  4
- Project 3-1 (**)  

Period 3.3  
- Project 3-1 (KEN3300)  6

Period 3.4  
- Data Analysis (KEN3450)  4
- Operations Research Case Studies (KEN3410)  4
- Intelligent Systems (KEN3430)  4

Period 3.5 to 3.6  
- Bachelor’s thesis (KEN3500)  18

* Third year students choose three electives per period out of the optional courses during period 1 and 2
** Project 3-1 will start in period 3.1 with weekly meetings. The credits for the project will become available at the end of period 3.3.

For each period, we will give a short explanation of the various courses. Before the start of each period, the students will receive detailed information about the content, the study material, the teaching form, the schedule, and the examination method.

Period 3.1

Semantic Web (optional course) (Code: KEN3140)

Examiner: Prof. dr. M. Dumontier and Dr. K. Moodley

Desired Prior Knowledge: Logic.

Description: Most of the information available on the World Wide Web (WWW) is not directly understandable for computers. For instance, web pages are designed for human readability. Computer programs have difficulty in interpreting the information presented on web pages. The focus on human readable information introduces restrictions on what computer programs can do to support human users in tasks such as:

- finding information
- buying goods
- making travel plans

The Semantic Web should eliminate these restrictions by separating the content of what is presented on a web page from the way it is presented. In recent years, the focus has shifted to providing data, independent of webpages (for example: Linked Open Data (LOD)) Ontologies are used to provide a shared conceptualization of information. Ontologies form the basis of the Semantic Web, Knowledge Based System, Databases, etc., and they play an important role in data exchange and interoperability in many domains. Ontologies are applied in the bio-medical domains, in data mining applications, in Linked Open Data (LOD), in websites based on semantic technology, etc.
Since ontologies are intended to be shared between different systems, defining an ontology is a challenging task.

This course will focus on the standards the World Wide Web Consortium (W3C) is defining in order to realize the Semantic Web. The course also addresses the underlying knowledge representation formalisms of the current semantic web standards. Moreover, the course will address the engineering principle of creating an ontology. Note that the course does not address standards for making websites.

Knowledge and understanding: Making the student familiar with the developments and standards of the Semantic Web. The student will get insights in semantic web standard, such as RDF, RDFa, SPARQL and OWL2. Moreover, the students will get some basic insight in the semantics of RDF and the Description Logic underlying OWL. Finally, the student will be made familiar with the ontology development process, and criteria for evaluating an ontology. The student should understand the role of upper ontologies and ontology design patterns, as well as the philosophical choices they represent.

Applying knowledge and understanding: The student should be able to build applications using semantic web standards such as RDF, RDFa, SPARQL and OWL2. The student should also be able to develop an ontology for an application domain.

Making judgements: The student should be able to judge whether and how semantic web standards can be applied in applications. The student should also be able to judge the quality of an ontology.

Communication: The student should have sufficient understanding of the Semantic Web and its standards in order to explain why and how an application should be set up using semantic web standards. The student should also be able to explain and defend the choices made in the ontology engineering process.

Learning skills: The student should be able to study the literature about semantic web developments.

Study material:

Recommended literature: The Semantic Web documents at the World Wide Web Consortium (W3C) web site.

Examination: Individual assignments (3*10%) and a final written exam (70%). Participation in the practical sessions is required for receiving a grade.

ECTS: 4

Game Theory (optional course) (Code: KEN3130)

Examiner: Prof. Dr. F. Thuijsman

Prerequisites: Discrete Mathematics.

Description: We introduce the field of Game Theory. Game Theory is the mathematical study of problems, called games, that involve two or more decision makers, called players, who each have their own individual preferences over the possible outcomes. In a game, each player always aims to maximize their individual payoff and chooses the actions accordingly. These actions may be probabilistic or deterministic, depending on the situation. Meanwhile the player reasons logically about actions that might be taken by the other players. A basic difference exists between strategic and non-strategic models. Both types of models and their solution concepts will be discussed. Issues like value, fairness, manipulations, threats, optimality and rationality will be addressed.
Knowledge and understanding: Students can recognize and classify the main types of games, i.e. cooperative games, strategic games, bipartite matching problems, and formulate the main solution concepts value, optimal strategies, Nash- and correlated equilibrium, as well as a number of algorithms to calculate these.

Applying the use of Knowledge and understanding: Students can calculate solutions for the different types of games

Making judgements: Students can explain advantages and disadvantages of different solution concepts. They are able to judge correctness of solutions presented

Communication: Students can explain and defend correctness of their solutions.

Learning skills: By the end of the course, students will be able to autonomously and critically reflect upon the pros and cons of different types of games for modelling competition and cooperation. This includes considerations on the computational aspects with respect to different solution concepts.

Study material: Lecture notes.

Examination: There will be a closed book written exam at the end of the course.

ECTS: 4

Prolog (optional course) (Code: KEN3234)

Examiner: Dr. J.W.H.M. Uiterwijk.

Prerequisites: Logic, Introduction to Computer Science 1.

Description: This course offers an introduction to the field of Artificial Intelligence (AI), based on one of the most important non-procedural programming languages: PROLOG. A detailed introduction into this declarative programming language is given. As an illustration, we will show how standard AI techniques can be applied in PROLOG and how expert systems can be realized with it. The course will be supported by a practical training, in which the acquired techniques can be put into practice in concrete PROLOG applications.

Knowledge and understanding: Students will be able to formulate problems in a logical context and to translate problem formulations into small PROLOG programs. Furthermore, students will have acquired active programming experience in PROLOG and are able to represent and manipulate data in knowledge-based systems.

Applying knowledge and understanding: Students will implement and test PROLOG programs and are able to use a suitable PROLOG environment for developing and testing programs. Furthermore, students can model problems in a logical way such that this model easily can be transformed into a PROLOG program. Students will also learn how typical PROLOG constructs, like recursive and functional programming, can be applied to other domains.

Making judgements: Students can analyze problems and decompose them in subproblems suitable for solving and can judge when functional or declarative programming languages seem more suited for their programming tasks than imperative languages.

Communication: program constructs must be easily understandable by peers and others experts.

Learning skills: Students will learn to model logical constructs independently and to critically reflect on one’s solutions.


Exam: Written exam; during the course the students will receive three practical assignments, that, if they receive a sufficient grade, may earn them up to a total of one bonus point.

ECTS: 4
Computer Security (optional course) (Code: KEN2560)

Examiner: Dr. A. Zarras
Tutor: Dr. A. Zarras

Desired Prior Knowledge: Introduction to Computer Science (1 & 2), Data Structures and Algorithms, Software Engineering, Databases

Description: Computer security is the process of detecting and preventing unauthorized and illicit access to a computer. As information systems have become mandatory in the commercial world, coupled with the increased frequency of security incidents, organizations now recognize the need for a comprehensive security strategy. The course will introduce a wide range of topics in computer security and online privacy. The main objective of the course is to cultivate a security mindset by discussing various attack techniques and defenses. Some of the topics we will explore include computer security technology and principles, software security and trusted systems, management issues, and Internet security.

Knowledge and understanding: Students will gain an in-depth understanding of computer security fundamentals and their application in real-world scenarios. In detail, they will understand the principles of a secure system and the potential attacks that can compromise it.

Making judgements: By understanding the fundamentals of computer security and by realizing their assignments, the students will be able to understand and avoid mistakes when designing a system. In principle, after completing this course, they will be able to design and develop secure systems on their own.

Communication: The class will consist of lectures in which several computer security issues will be discussed. In parallel, there will be bonus assignments where the students will have to solve some of the most important issues we discussed in the classroom. This way they will be able to learn all the described principles in depth.

Learning skills: Students will be able to design and implement secure systems, avoid common mistakes that can introduce vulnerabilities to their systems, and work on real-world security problems.

Recommended literature:

Exam: Assignments and Project
ECTS: 4

Robotics and Embedded Systems (optional course) (Code: KEN3236)

Examiner: Dr. R. Möckel.

Desired Prior Knowledge: Calculus, Linear Algebra, Machine Learning.

Prerequisites: Introduction to Computer Science 1 and 2.

Description: Nowadays, a variety of products require that algorithms from data science and artificial intelligence are adapted to and implemented in robotic and embedded systems. Applications that heavily rely on intelligent robotic and embedded systems include self-driving cars, autonomous drones, intelligent industrial robots in (semi-) autonomous factories, smart phones, intelligent medical devices, and distributed intelligent embedded devices in smart homes.

In this course, students receive an introduction to the fields of robotics, embedded systems, and real-time control. Students obtain an overview of state-of-the-art intelligent robotic and embedded systems in academia and industries. Students gain hands-on experience in programming embedded robotic systems using embedded processors and a modular robotic system developed at DKE. Students learn about communication standards for embedded systems, sensors, and actuators.
Student practise and strengthen their expertise in data science and Artificial Intelligence by applying mathematical methods for controlling robotic systems: They study control techniques including PID control, forward and inverse kinematics as well as locomotion control and learning using central pattern generators. The course concludes with a robot competition where students build and program robots using a modular robotic system.

**Knowledge and understanding:** Students obtain knowledge in designing, building, and programming robotic and embedded systems. Students learn how to apply mathematical concepts like dynamic systems for controlling robotic systems in real-time. Students further obtain knowledge about sensors and motor control and study the application of machine learning and mathematical methods for learning and optimizing control parameters. Students receive training in the programming language C - the most popular languages for programming microcontrollers.

**Applying knowledge and understanding:** After successful completion of this course, students can analyse, apply, implement, and validate control techniques in embedded and robotic systems with and without real-time constraints. Students can apply techniques from machine learning, search, and optimisation to obtain parameters for embedded control systems as required in many professional academic and industrial applications.

**Making judgements:** Students learn to judge where real-time systems are required and embedded systems can be beneficial. Students further learn to critically analyse the use of robotic systems in a variety of scenarios and to make design choices for robotic and embedded systems. By introducing students to a variety of state-of-the-art robotic systems, the course lays the foundation so that students can process professional literature in robotics and embedded systems.

**Communication:** Students will be able to 1) discuss robotic and embedded systems professionally and critically, 2) plan, discuss, implement, and validate projects in robotics and embedded, 3) present the results of project assignments in form of video, and to 4) critically analyse and explain control techniques for robotic systems to a general and professional audience.

**Learning skills:** Students are able to autonomously and critically reflect upon the abilities and limitations of robotic and embedded systems in order to keep up with new developments in the field. Students can further assess the capabilities and limitations of their own solutions to a control or machine learning problem in robotics, and to identify follow-up literature, which goes beyond the scope of the material presented in the course.

**Study material:** Course material will be provided during the lectures.

**Exam:** The final course grade is 80% of the final written “closed-book” exam grade plus 20% of the assignments grade.

**ECTS:** 4

**Digital Society (optional course) (Code: KEN3111)**

**Examiner:** Dr. K. Wenz

**Prerequisites:** none.

**Desired Prior Knowledge:** none.

**Description:** Digitalization has a profound impact on our society. We can observe changes in different areas. What digital technologies do, what they look like and how they relate to each other is not identical worldwide, but dependent on local practices as well. Usually new technologies are understood as innovation and progress: and indeed, digital technologies improve a broad range of domains, such as healthcare or education. New possibilities as e.g. participation in our digital cultures arise but also...
new inequalities, as the access and competences needed for participation are not evenly distributed and the platforms that allow for participation also harbour new mechanisms of control and surveillance. The pace and diversity of these developments ask for continuous investigation and reflection. It requires work to shape and use technologies in ways that contribute to the public good. Moreover, digital technologies have also led to highly problematic developments such as electoral manipulation, fake news and algorithmic discrimination. Technological developments are often conceived as predefined or given. Does a society’s technology drive the development of its social structure and cultural values? Scholars in science and technology studies have shown that technology and society are deeply intertwined. Technology is inherently social. Technologies are shaped by people; they emerge and are embedded in social practices.

The aim of this course is to investigate the consequences of digitalization for our society/societies. These consequences have been differently valuated: participation vs. exploitation of users, innovation as enhancement vs. challenge, ethics and technomoral change vs./and sustainability. We will discuss digitalization from
- a social perspective when we read about digital participation and how technology and society are intertwined
- a political perspective when we discuss activism, digital citizenship but also problems of manipulation and verification (as in the case of fake news and deep fakes)
- a cultural perspective when we analyze imaginaries and discourses around innovation of technology and promises being made
- a legal perspective when we discuss privacy and the attempts to adapt privacy laws
- an ethical perspective when we discuss design decisions, privacy but also technomoral change and questions of environment and sustainability.

The course is structured in the following way:

Transformations
(digital participation, digital citizenship, data-activism)

Imaginaries
(innovation and technomoral change)

Disruptions
(fake news and deep fakes, sustainability and e-trash)

Knowledge and understanding: Student acquire knowledge on the impact of digitalization on society.

Applying knowledge and understanding: Students learn to understand the interrelation between digital technology and sociality.

Making judgements: Upon completion of the course, students can reflect on ethical challenges related to digitalization.

Communication: Students are able to communicate central topics related to digitalization to an audience of non-IT-experts (e.g. the debate will bring students from the FASoS BA Digital Society and DKE students following this course together to train both groups to communicate topics related to digitalization from a social science and IT perspective.)

Learning skills: Students have acquired the learning skills to reflect critically in written form on a topic related to the digital society but also to do so orally in a presentation and debate.

Study material: The literature will be made accessible via the reference list of the library.

Exam: (group) presentation in class (1-3 students) per task (25% of the final grade), 2 short academic papers of 1500 words each (2x25% of the final grade) and participation in a final debate (25% of the final grade).
The resit of the different parts will be as follows:
If a resit is needed for the (group) presentation the presentation will be given via video (e.g. Zoom or Skype)
If a resit is needed for the papers they can be rewritten and improved based on the comments of your tutor
If a resit is needed for the debate (in case a student does not show or participate in the debate) the student can write a 1500-word paper on the content of the debate instead.
ECTS: 4

**Period 3.2**

**Large Scale IT and Cloud Computing (optional course) (Code: KEN3239)**
**Examiner:** Dr. Thomas Eifert, Bastian Küppers, M.Sc.
**Desired Prior Knowledge:** Introduction to Computer Science 1, Databases
**Prerequisites:** none
**Description:** The course offers a comprehensive introduction to the field of scalable IT systems, so-called “Big IT”, and cloud computing. After a technical introduction to the available methodologies of setting up and running scalable systems, use cases are presented. These use cases emphasize the correlation of the processes and requirements of large institutions and possible technical solutions. A special focus is put upon the question which technological platform is best used for which use case as well as process aspects of scaling. Security aspects specific to cloud computing are discussed along the use cases. Cloud computing, as a special case of scalable IT, is discussed in detail. Different cloud providers are presented and evaluated in the context of university requirements, i.e. requirements posed by research and teaching processes.
**Knowledge and understanding:** Students acquire an overview of existing technologies for scalable systems, and specific security requirements for the different use cases.
**Applying knowledge and understanding:** Students are able to understand scalability and are able to set up and use a scalable IT system. In addition, students are able to evaluate high scalable IT solutions in terms of benefits and security risks.
**Making judgements:** Students are able to analyze the requirements of a specific use case and can decide which technology is best used for that case of application.
**Communication:** students are able to communicate about scalable IT systems and specific security requirements.
**Learning skills:** Additionally, students are able to analyse the interdependencies between large organizations, processes and IT solutions - taking into account security-related aspects - and to design suitable solutions using cloud offerings.
**Study material:** Lecture notes
**Recommended literature:** TBA
**Exam:** Assignments and Project
**ECTS:** 4

**Logic for Artificial Intelligence (optional course) (Code: KEN3231)**
**Examiner:** Dr. ir. ing. N. Roos.
**Desired Prior Knowledge:** Knowledge of propositional and predicate logic.
**Prerequisites:** Logic.
**Description:** Logics form the formal foundation of knowledge representation and reasoning, which is a fundamental topic in Artificial Intelligence. Logics play a role as an analysis aid and as a knowledge-representation formalism. Moreover, the
semantics of logics enables us to evaluate the intended meanings of knowledge representation formalisms, and the correctness and completeness of reasoning processes.

Humans make assumptions in their day-to-day reasoning. Examples of reasoning with assumptions are common sense reasoning, model-based diagnosis, legal argumentation, agent communication and negotiation, and so on and so forth. The assumptions humans use in their reasoning may be incorrect in the light of new information. This implies that conclusions may have to be withdrawn in the light of new information. Therefore, this form of reasoning is called non-monotonic reasoning and the underlying logics are called non-monotonic logics.

The course will cover model-based diagnosis as an application of reasoning with assumption, standard logics extended with defeasible rules, argumentation systems, and the semantics of reasoning with assumptions and defeasible rules, and closure properties of the reasoning systems.

Knowledge and understanding:
- The student should be able to describe non-monotonic logics and argumentation systems.
- The student should be able to identify the logic underlying specific forms of knowledge representation.
- The student should be able to describe and discuss the semantic of non-monotonic logics.

Applying knowledge and understanding:
- The student should be able analyze important properties of practical formalisms for knowledge representation and reasoning.
- The student be able to apply non-monotonic logics and argumentation systems to practical problems

Making judgements:
- The student should be able to judge whether specific knowledge representation formalisms are able to represent the intended meaning of the knowledge to be represented.
- The student should be able to analyze whether conclusions derived from a knowledge representation are correct and complete.

Communication:
- The student should be able to explain how logic can be used as a tool for analyzing a knowledge representation problem.
- The student should be able to explain issues involved in the handling assumptions in a knowledge-representation.

Learning skills:
- The student should be able to study autonomously the literature describing the applications of logics for knowledge representation and reasoning.

Study material: Syllabi.

Recommended literature: A syllabus and scientific literature.

Examination: Written exam at the end of the course. A bonus of 1.0 point can be earned by a series of bonus assignments.

ECTS: 4

Parallel Programming (optional course) (Code: KEN3235)
Examiner: Prof. dr. Hans Pflug, Bastian Küppers M.Sc.
Prerequisites: Introduction to Computer Science 1 and 2, Data Structures and Algorithms.
Description: Parallel programming introduces the students to the paradigm of parallel computing on a computer. Nowadays almost all computer systems include so-called
multi-core chips. Hence, in order to exploit the full performance of such systems one needs to employ parallel programming. This course covers shared-memory parallelization with OpenMP and Java-Threads as well as parallelization with message passing on distributed-memory architectures with MPI. The course starts with a recap of the programming language C followed by a brief theoretical introduction to parallel computing. Next, the course treats theoretical aspects like MPI communication, race conditions, deadlocks, efficiency as well as the problem of serialization. This course is accompanied by practical labs in which the students have the opportunity to apply the newly acquired concepts. After completing this course students will be able to write parallel programs with MPI and OpenMP on a basic level, and deal with any difficulties they may encounter.

Knowledge and understanding: Students recall the basic concepts for parallel programming and recognize important parallelization patterns.

Applying knowledge and understanding: Students are able to write parallel software code using MPI, OpenMP, and Java Threads.

Communication: Students are able to explain why a specific pattern is adequate for a given problem.

Learning skills: Students are able to study autonomously the literature describing parallel programming in order to comprehend important details and problems of the field.

Study material: Course notes and several codes will be provided online.

Recommended literature: Parallel programming with MPI; Peter Pacheco; Morgan Kaufmann (1996); (a very early revision is available online)

Exam: Written exam.

ECTS: 4

Software and Systems Verification (optional course) (Code: KEN3150)

Examiner: Dr. Pieter Collins
Tutor(s): Dr. Pieter Collins

Desired Prior Knowledge: Reasoning Techniques, Theoretical Computer Science

Description: Have you ever written a program with a bug in it? Then this course is for you! Software verification tools can check whether your program works by showing that it correctly satisfies its specification, or finds a case in which it can go wrong. Unlike unit testing and other software validation methods, verification tools use formal methods to rigorously prove correctness. Similar techniques can be used to show that (mathematical models of) cyber-physical systems work as designed. In this course, we will start by and introducing the main notions of object-oriented program verification, including pre- and post-conditions for methods, and class invariants. We shall use Hoare logic to convert programs and their specifications into logical statements to be proved. We shall apply these techniques to the verification of simple programs written in Java.

In the second part of the course, we consider formal models of software and systems as labelled transition systems (automata), using temporal logics for specification, and consider the fundamental algorithms for verification. We shall apply these algorithms to simple discrete verification problems, such as vending machines and communications systems, modelled using a specification language such as SMV. Finally, we will look at simple continuous systems, such as robots and electronic systems, and show how to verify these using rigorous numerical methods based on interval arithmetic.

Knowledge and understanding: By the end of the course, students are able to:
Recognise the difference between formal verification and validation. Explain the
various kinds of annotations used in program specification. State the deduction and precondition rules of Hoare logic. Interpret linear temporal logic formulae. Distinguish rigorous numerical methods, notably how interval arithmetic differs from floating-point.

**Applying knowledge and understanding:** Students are able to write formal specifications for simple programs. Furthermore, students can use Hoare logic to reduce program specification to first-order logic statements, and justify these. Students are able to construct Büchi automata accepting temporal logic formulae and can apply interval and affine arithmetic for verifying properties of continuous systems. Moreover, students are able to write annotations for object-oriented software, use software for model-checking discrete systems and use software for rigorous numerics to verify safety of simple continuous systems.

**Making judgements:** Students are able to determine the most appropriate modelling framework and verification tools for a given problem.

**Communication:** Students can write and read formal specifications and can discuss informal design goals and their translation into formal specifications.

**Learning skills:** Students will critically reflect on their own human reasoning and the potential of digital computers.

**Study material:** Course notes.


**Exam:** Written exam (100%)

**ECTS:** 4

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**Introduction to Bio-Informatics (optional course) (Code: KEN3440)**

**Examiner:** Dr. R. Cavill

**Desired Prior Knowledge:** Introduction to Computer Science 1, MatLab.

**Prerequisites:** None.

**Description:** This course presents a general introduction to the fundamental methods and techniques of bioinformatics in biomedical and biological research. The objective is that the students will acquire a general understanding of bioinformatics methods at the algorithmic level and will therefore be able to read and understand publications in this field, and – to some extent – apply their knowledge to concrete biological problems. This relates to the major areas of bioinformatics like sequence alignment, phylogenetic analysis, gene finding, and gene expression analysis. This course consists of a series of closely related lectures and computer classes, based on relevant real case studies. In the lectures, the main theoretical aspects are presented. In the computer practicals, the students work to analyse real data using the techniques they have encountered. By extensively exploring the case study, the students acquire a thorough understanding about the subject.

**Knowledge and understanding:** Students should be able to perform common analyses on both sequence data and numeric data from omics experiments. This includes sequence alignment, building phylogenetic trees, applying hidden Markov models, detecting differentially expression and performing pathway analysis.

**Applying knowledge and understanding:** For all the above topics students should be able to demonstrate the algorithms on paper with simple examples and apply the algorithms appropriately on realistic datasets using a computer.

**Making judgements:** After successful completion of the course, students will be able to judge the use, quality, and correctness of different bioinformatics algorithms and results.
Communication: After this course, students will be able to explain the algorithmic bases of bioinformatics problems and interact with biologists to provide recommendations of analysis approaches in the situations studied.

Learning skills: After successful completion of the course, students will be able to independently read bioinformatics literature to further their knowledge.


Exam: Written exam (50%) + assignments (50%).

ECTS: 4

Introduction to Quantum Computing (optional course) (Code: KEN3241)
Examiner: Dr. Georgios Stamoulis.

Prerequisites: Linear Algebra.


Description: This course offers an introduction to the interdisciplinary field of quantum computation. The focus will lie on an accessible introduction to the elementary concepts of quantum mechanics, followed by introducing the mathematical formalism and a comparison between computer science and information science in the quantum domain. The theoretical capability of quantum computers will be illustrated by analysing fundamental algorithms of quantum computation and its potential applications.

Quantum technology has become one of the most prominent interdisciplinary fields of recent research. This course will focus on introducing the mathematical concepts underpinning quantum computation, and on explaining how this new computational paradigm might potentially offer possibilities beyond the scope of conventional computers. Topics that will be introduced and discussed include: (i) most common models of quantum computation (e.g., quantum circuits and measurement-based quantum computing). (ii) An exposition of the machinery borrowed from quantum mechanics, such as superposition of states, quantum entanglement, (de)coherence etc., which gives rise to the potential speed-up of quantum algorithms over their classical analogs. (iii) Some of the most common quantum algorithms (searching, factoring etc.) and protocols (quantum teleportation, EPR paradox). The course will finish with an exposition of potential applications of quantum computation and algorithms in other fields (such as security/cryptography, AI, optimization etc.)

Important: no prior knowledge in quantum mechanics is assumed or required, and all necessary concepts will be introduced and motivated from a mathematical and theoretical computer science point of view. Possible quantum architectures and/or related hardware issues will not be discussed.

Knowledge and understanding: By the end of this course, students are able to understand the differences between classical and quantum computation: Where is the computational power of quantum machines coming from? What are the limits of this new computational paradigm? What does the term “quantum supremacy” mean and why it is important? How likely is it ever to be achieved and what would it mean for our current understanding of the computational landscape?

Applying knowledge and understanding: Students are able to understand some of the most famous quantum algorithms, and to demonstrate where their power comes from. They will be able to judge how this potential computational power can be leveraged, and how it can be applied to other fields in a beneficial way.

After successful completion of this course, students are able to understand and use the mathematical framework of quantum computing to solve computational problems.
Making judgements: Students are able to judge and identify the settings where the potential quantum power might be beneficial and how they can leverage this. Students will further be able to analyse simple quantum algorithms for different computational problems.

Communication: Students are able to discuss quantum computation critically and judge not only its benefits but, equally important, its shortcomings. During lectures and practical assignments, students will be exposed to a different way of thinking about computation that will also enhance their understanding on classical computation.

Learning skills: Students are able to critically read and understand scientific papers on quantum computing. To explain and analyse quantum algorithms described in quantum circuit or measurement-based quantum computing models. Finally, to relate quantum complexity classes to the classical ones.

Recommended Study material:

Course material will be also provided during the lectures.

Exam: The final course grade is 100% of the final written “closed-book” exam grade.

ECTS: 4

Period 3.3

Project 3–1 (Code: KEN3300)
Examiners: Dr. R. Möckel, Dr. A. Briassouli, Dr. Anna Wilbik
Tutor(s): Dr. R. Möckel and Dr. Dr. K. Schüller
Prerequisites: Project 2-1.
Description: Project 3-1 consists of two distinct tracks: a research project at DKE and a professional project proposed by companies affiliated with BISS (Brightlands Institute for Smart Society). In the first week of period 1, students indicate their preference by ranking these projects. Groups are created by means of an algorithm that minimizes regret, and allocates students to their most preferred options.

About the DKE project. Students work in small groups, guided by lecturers of the subjects concerned and by the tutors. During the project, they apply their knowledge in data science, knowledge engineering, and artificial intelligence to robotic systems. Depending on their chosen specialization within their project group, students study and search for solutions in at least one, typically in multiple of the following fields: control, computer vision, signal processing, artificial intelligence, modelling and simulation of physical systems. Students learn and practise the application of autonomous adaptation and closed-loop control to obtain robust solutions that can be successfully operated under real-world constraints.

During the BISS project, students participate in small groups and receive guidance from a tutor, a teacher with knowledge of the subjects concerned, and a content expert from the company. Students learn how to apply their knowledge in data science, knowledge engineering, and artificial intelligence to solve real-world problem that arise in a professional environment, and how to interact with a client from the industry.

Project Skills period 3.1 & 3.2:
Group CV Check (online or on-site): during this class, you will receive tips and feedback on how to write a professional résumé (i.e. Curriculum Vitae).
Networking skills (online or on-site): during this class, you will learn hands-on tips to build an interesting network to support you in your search for a job or internship. Both of the above classes are provided by instructors of the UM Career Services.

**Study material:** Period book 3.1-3.3. Maastricht University, DKE.

**Exam:** The project will be assessed based on report, product, and presentation.

**ECTS:** 6

### Period 3.4

**Data Analysis (Code KEN3450)**

**Examiner:** Dr. G. Spanakis

**Tutor(s):** None.

**Desired Prior Knowledge:** Calculus, Linear Algebra, Mathematical Modelling & Simulation, Machine Learning, Introduction to Computer Science 1 and 2.

**Prerequisites:** Calculus

**Description:** This course aims at preparing students on how to be a successful “data scientist”. The crucial processes of inspecting, cleaning, transforming, restoring and preparing data for modelling are tackled. Different types of data are going to be explored through case studies (“clinics”) that a modern “data scientist” has to deal with. Furthermore, several techniques from machine learning and mathematical modelling (multiple regression, classification, tree-based models, dimensionality reduction, etc.) are presented from the data analysis perspective and students learn how to apply these techniques to different types of data. Finally, the cornerstone of data analysis is presented: correct communication of the analysis outcome (storytelling, visualization, etc.).

**Knowledge and understanding:** Students are able to illustrate and explain data analysis and machine learning techniques with emphasis on modelling, and to give examples of different domains where data analysis can be applied

**Applying knowledge and understanding:** Students are able to examine datasets using techniques learned in course, and to experiment with different techniques for data modelling

**Making judgements:** After successful completion of the course, students are able to 1) judge the quality of data (of any kind), 3) to justify and rank which techniques should be applied in each problem and 3) to assess results of data analysis process

**Communication:** Students are able to present the results of different stages of data analysis to specialists and non-specialists and are able to decide on the correct communication medium (scientific, verbal and visual) of the analysis outcome

**Learning skills:** After successful completion of the course, students are able to suggest options for tackling different datasets combining verbal, numerical/scientific and visual descriptions, also taking into account the context cases (e.g. business, academic) or the domain of application. Furthermore, students are able to formulate data descriptions based on their characteristics and can suggest options for modelling data, doing basic frequency analysis and dimensionality reduction

**Study material:** Handouts.

**Recommended literature:** Selected chapters from the following textbooks:
A. Downey, Think Stats: Exploratory Data Analysis
James, G., Witten, D., Hastie, T., Tibshirani: An Introduction to Statistical Learning (with Applications in R)
J. Vanderplans, Data Science Handbook
S. Skiena, The Data Science Design Manual
J W. McKinney, Python for Data Analysis
Operations Research Case Studies (Code: KEN3410)
Examiner: Dr. S. Kelk and Prof. dr. F. Thuijsman.
Desired Prior Knowledge: Linear Programming.
Prerequisites: None.
Description: Operations Research (OR) is concerned with the best way to assign scarce resources to competing activities. It is for this reason an important branch of mathematics that is widely used in industry to support economically efficient decision making, but also in other application areas where discrete or stochastic optimization has a central role. In this course we will explore a number of themes both within deterministic OR (where all the problem data is known at the beginning) and stochastic OR (decision problems involving uncertainty and randomness). Themes within deterministic OR include the network simplex method (used for solving minimum-cost flow problems), integer linear programming and non-linear programming. Stochastic themes include queuing systems, Markov chains and Markov decision problems. As background students will be introduced to the methodological similarities and differences between OR and data science.
Knowledge and understanding: Students can recognize, classify and distinguish some of the major types of OR models, i.e. transportation and network optimization models, integer and non-linear programming, Markov chains and Markov decision problems, queueing models.
Applying knowledge and understanding: Students can apply a wide variety of algorithms to calculate solutions for problems of the types mentioned above. Students will be able to translate simple real-world/industrial optimization problems into a format suitable for (variously) the transportation simplex, network simplex and integer linear programming.
Making judgements: Students can explain advantages and disadvantages of different models and algorithms. They are able to judge the correctness of solutions presented.
Communication: Students can explain and defend their solution methods.
Learning skills: Students will be able to critically reflect upon the scope and limitations of the learned models, and be able to identify follow-up literature describing paradigms, models and algorithms that go beyond the scope of the course.
Recommended literature: None.
Exam: Written exam, worth 100% of the credit.
ECTS: 4

Intelligent Systems (Code: KEN3430)
Examiner: Dr. ir. K. Driessens
Desired Prior Knowledge: Machine Learning
Prerequisites: None.
Description: The course offers an introduction to intelligent systems. Based on the metaphor of a computational agent (that is, a software program or a robot which acts and interacts flexibly and autonomously in order to achieve some goal), basic concepts and methods from agent technology are discussed. Topics covered are the
concept of artificial intelligence, characteristics of an agent and agent architectures, agent cooperation and competition among agents, behaviour-learning agents in a multi-agent environment. An emphasis is made on the complexity of interacting agents, but also subsystems within a single agent. In the practical part of the course, the students build up their experience with the implementation of a number of different types of agents. Some topics of this course are discussed more extensively in the master's course Multi-Agent Systems and Advanced Concepts in Machine Learning.

**Knowledge and understanding:** Students are able to compare and discuss benefits and drawbacks of a number of different agent technologies. They can also explain the complexities arising from interactions between multiple techniques within a single agent, and the interactions between agents and systems.

**Applying knowledge and understanding:** Students will be able to implement of a number of different types of agents architectures and agent-subsystems.

**Making judgements:** The student will be able to judge whether it is beneficial to use agent technology over other approaches for handling a given problem, and which agent architecture might fit best.

**Communication:** The student will gain a working knowledge of agent terminology and will learn to motivate his/her choices concerning the application of agent technology.

**Learning skills:** Students have to reflect upon their knowledge and recognize the need for continued learning as they are confronted with the complexities involved with applying the knowledge gained in their bachelor studies and linking individual techniques into a working system.

**Study material:** Course slides; supplementary material consisting of research papers and book chapters.

**Recommended literature:**

**Examination:** Written exam (80%) + assignments (20%).

**ECTS:** 4

**Period 3.4 to 3.6**

**Bachelor's Thesis (Code: KEN3500)**

**ECTS:** 18

**Bachelor's thesis Data Science and Artificial Intelligence**

At the end of the Bachelor's study in Data Science and Artificial Intelligence each individual student has to write a thesis. This thesis has to be designed as a scientific paper of 8 to 10 pages using a standard (LaTeX) design. Students are expected to conduct a pro-active and independent research on their topics. This includes the search and reading of related work. The topics must be discussed with the potential thesis supervisor(s) and a research plan must be submitted to and approved by the Board of Examiners as an initial step. The thesis has to be accompanied by relevant attachments and software. Students will present the thesis in a conference. This means that a strict submission form will be used. In order to start working on the thesis, a student needs to have obtained at least 140 ECTS (among which are 60 ECTS of the first year, and 40 ECTS of the second year).
General procedure
below is an indication for these phases. A special bachelor’s thesis coordinator will supervise the procedure and schedule.

November
Phase 0: Thesis Topic meeting
Potential topics and research fields will be presented by staff.

January
Phase 1: Topic selection
During the skills class, each student selects a topic (and problem statement). A thesis supervisor with experience in the field of choice is appointed to each student. Every student hands in a signed bachelor’s project plan to the bachelor’s thesis coordinator. A special bachelor’s thesis coordinator will supervise the phase.

Periods 3.4–5: February - May
Phase 2: Research
In this period students conduct their own research. This will preferably be guided in groups by the thesis supervisor. Further two seminars will be organised during which the students present their progress. A special bachelor’s thesis coordinator will supervise the phase.

Phase 3: Writing
Parallel to the research, a scientific article is written.

Period 3.6: June
In Period 6, the research is finished and the first version of the thesis is expected (first submission). The thesis supervisor will evaluate the paper (or make sure this happens) and gives a first reaction within around a week. A second assessor will also evaluate the paper during this week. The second and final submission will take place at the end of the second week of period 3.6 (concrete dates will be announced).

Phase 4: Preparation for presentation
In the second week of period 3.6, the preparation for the final presentation will start for every student individually. The presentations will be accompanied with slides created with PowerPoint (or similar) and have a maximum length of 10 minutes.

Phase 5: Presentation
The bachelor’s theses will be presented in the third week of period 3.6 in a scientific conference setting. The presentations have a maximum length of 15 minutes per student (including questions). The conference is open for all students and employees from Data Science and Artificial Intelligence and anyone else who might be interested. The final decision on the grade for the bachelor’s thesis will be made shortly after the presentations. A special bachelor’s thesis coordinator will supervise the phase.

Re-sit: In case the student fails to present his/her work at the Bachelor conference, the student gets one opportunity to defend his/her work at the next bachelor conference. If the student does not participate at or pass any of those two conferences, the student has to select a new topic and submit a new thesis plan. For students not finishing at the June Conference, there is one re-sit possibility in a conference setting at the end of August.

Requirements for the bachelor’s thesis project
For the bachelor’s thesis, every student has to conduct a short scientific research project. This can be an empirical as well as a theoretical research. The topic for the research
The department will offer a list of potential research topics. The topic and the research questions have to be approved by the thesis supervisor. To achieve this, the student will create a bachelor’s project plan using the form provided by the Board of Examiners which contains the following:

- Date;
- Name of the thesis supervisor;
- Name of the second examiner;
- Title of the bachelor’s thesis, start- and end date of the thesis project and a planning;
- Short description of the research question;

This plan will be signed by the student, the thesis supervisor and the second examiner and then handed in to the bachelor’s thesis coordinator. It is possible to execute the bachelor thesis project as an external training period. This should be well defined in the bachelor’s thesis plan. In this case, the plan should also include the name of the company, the name of the external supervisor, the size of the project and any agreements about compensation and confidentiality. The plan should also be signed by the external supervisor. The external research will not start before period 3.5. The research needs to be original in such a way that the thesis supervisor is convinced that this research has not been done before. The research also needs enough depth and still it must be possible to finish it in the set amount of time. It is possible for multiple students to cooperate in a research project as long as it is clear who did what. Moreover, every student has to write their own thesis reflecting their part of the research.

Requirements for the bachelor’s thesis document

Content aspects

The thesis describes the cause, research question, approach and results of the research. This has to be done in a clear, structured and scientific manner which includes:

- a clear introduction in which the cause and research questions are presented;
- a clear conclusion, based solely on the already used thought out principles and derived results;
- a clear line is shown between problem statements approach methods and the derived results;
- a motivation of the followed approach;
- an adequate description of the followed approach;
- an honest, clear and concise description of the derived results, if necessary using tables;
- a discussion of the results;
- the usage of relevant and recent literature;
- the correct usage of references;
- the adequate usage of the literature for the reasoning in the thesis.

Design aspects

The number of pages of the thesis is between 8 and 10 (no more, no less) in the designated LaTeX format, including images and references. This thesis should at least contain:

- title;
- author;
- abstract;
- one or two keywords;
• list of references;
• page numbers.

It goes without saying that the correct, scientific references are used for used resources (by using the designated BiBTeX reference style). Images and tables are accompanied by an index and caption. Mathematical formula, definitions, etc. have to be properly designed and numbered. The start and end of mathematical formula have to be properly defined.

Language aspects
The thesis has to be written in Dutch or English, considering correct spelling, syntactical structure of sentences and structure of content in paragraphs. The target audience is fellow Data Science & Artificial Intelligence students. Any jargon and/or abbreviations have to be explained unless they are common knowledge for this audience (e.g. CPU).

Citations
It is allowed to use several short citations with a maximum length of two sentences. These citations have to be clearly referenced and have to be typographically distinguishable (that is, citations are placed in quotes). Non-allowed citations or missing references will result in an unsuccessful result.

Assessment
The assessment will take place based on the contents and design of the thesis and the presentation of this thesis.
2.4 Master’s Programmes Artificial Intelligence and Data Science for Decision Making

2.4.1 Master’s Artificial Intelligence

General introduction
The Master in Artificial Intelligence (AI) is a two-year advanced programme organised by the Department of Data Science and Knowledge Engineering. The focus of this programme is on the understanding, design and creation of intelligent systems, such as those used in robotic systems, games or artificial personal assistants. Artificial Intelligence has become a very active domain in both academia and industry. It has given rise to computer programmes and robots that learn from experience, recognise and adapt to patterns in their environment, and reason strategically in complex decision-making situations.

The impact of the field of Artificial Intelligence is pertinent due to the key role it plays in technological applications that have become indispensable in society, such as simple personal assistants that adapt the settings of your smart phone to automatically recognised activities (e.g. driving or attending a meeting, automated trading software used in real markets to respond to rapid price changes, interactive computer games that include human like opponents, robotic assistance in the exploration of dangerous environments, etc.). In this master’s programme, you are trained to become an expert and capable of dealing with todays and future challenges in the field of Artificial Intelligence.

The master’s programme Artificial Intelligence covers a range of subjects emphasizing the following research topics as its core:
1) Intelligent techniques for playing and solving (board) games and controlling virtual characters in video games;
2) Situated agents to study the control and coordination of embodied agents, i.e. robots (e.g. autonomous flying robot swarms);
3) Multi-agent systems of collaborating autonomous intelligent systems;
4) Formal techniques for reasoning in agents and representing and communicating knowledge;
5) Machine learning to extract useful patterns and knowledge from experience and make predictions about the future;

The members of the teaching staff are actively involved in one or more of these research topics. As a result, the educational contents of the courses relate directly to the research performed.

2.4.2 Master’s Data Science for Decision Making

General introduction
Data Science for Decision Making is the science of making informed decisions. It has widespread applications in business and engineering. In today’s world, many companies and organisations collect all sorts of data in large amounts. They aim to extract useful information from it, to recognize patterns and anomalies. Data Science for Decision Making provides the mathematical tools to analyse and model these big data. It also provides and uses the computational software that is the key to data science.
The two-year master’s programme in Data Science for Decision Making teaches the use of applied mathematics to analyse and optimize processes, problems and operations. Examples of applications are: discovering patterns in data such as images and time series, scheduling customer service agents, optimising supply chains, controlling dynamical systems, modelling biological processes, finding optimal strategies in negotiation, and extracting meaningful components from brain signals.

The master’s programme Data Science for Decision Making covers a wide range of research topics, focusing on the following ones in its core:

1) Data mining to extract useful patterns and knowledge from large data repositories;
2) Mathematical modelling and parameter estimation from data, system identification, model approximation and reduction of model complexity;
3) Algorithm design and analysis to efficiently deal with the challenges that the ever-growing amount of data pose;
4) Mathematics and algorithms associated with modelling and solving planning/scheduling problems.

The members of the teaching staff are actively involved in one or more of the research topics. As a result, the educational contents of the courses relate directly to the research performed.
Programme master’s AI Year 1

**Period 1**
- Foundations of Agents (KEN4115) 6 ECTS
- Intelligent Search & Games (KEN4123) 6
- Research Project AI 1 (**)

**Period 2**
- Multi-Agent Systems (KEN4111) 6 ECTS
- Advanced Concepts in Machine Learning (KEN4154) 6
- Research Project AI 1 (**)

**Period 3**
- Research Project AI 1 (KEN4130) 6

**Period 4**
- Autonomous Robotic Systems (KEN4114) 6 ECTS
  1 Elective Course from the following set:
  - Algorithms for Big Data (KEN4254) 6
  - Dynamic Game Theory (KEN4251) 6
  - Computational Statistics (KEN4258) 6
  - Advanced Natural Language Processing (KEN4259) 6
  - Research Project AI 2 (**)

**Period 5**
  2 Elective Courses from the following set:
  - Information Retrieval and Text Mining (KEN4153) 6 ECTS
  - Deep Learning (KEN 4257) 6
  - Planning and Scheduling (KEN4253) 6
  - Computer Vision (KEN4255) 6
  - Research Project AI 2 (**)

**Period 6**
- Research Project AI 2 (KEN4131) 6

Programme master’s AI Year 2

**Period 1, 2, 3**
- Electives * 30 ECTS

**Period 4, 5, 6**
- Master’s thesis AI (KEN4160) 30

Programme master’s DSDM Year 1

**Period 1.1**
- Data Mining (KEN4113) 6 ECTS
  1 Elective Course from the following set:
  - Signal and Image Processing (KEN4222) 6
  - Mathematical Optimization (KEN4211) 6
  - Stochastic Decision-Making (KEN4221) 6
  - Research Project 1 (**)

**Period 2**
- Model Identification and Data Fitting (KEN4242) 6 ECTS
  1 Elective Course from the following set:
  - Advanced Concepts in Machine Learning (KEN4154) 6
  - Applications of Image and Video Processing (KEN4244) 6
  - Information Security *with reservation (KEN4245) 6
  - Research Project 1 (**)

**Period 3**
- Research Project 1 (KEN4230) 6

**Period 4**
- Algorithms for Big Data (KEN4254) 6 ECTS
  1 Elective Course from the following set:
  - Building and Mining Knowledge Graphs (KEN4256) 6
  - Dynamic Game Theory (KEN4251) 6
  - Computational Statistics (KEN4258) 6
  - Advanced Natural Language Processing (KEN4259) 6
  - Research Project 2 (**)

**Period 5**
- Planning and Scheduling (KEN4253) 6 ECTS
  1 Elective Course from the following set:
  - Symbolic Computation and Control (KEN4252) 6
Deep Learning (KEN4257) 6
Information Retrieval and Text Mining (KEN4153) 6
Computer Vision (KEN4255) 6
Algorithms for Data Visualization (KEN4213) 6
Research Project 2 (**)

Period 6
Research Project 2 (KEN4231) 6

Programme master’s DSDM Year 2
Period 1, 2, 3 Electives * 30
Period 4, 5, 6 Master’s thesis DSDM (KEN4260) 30

* Note: during the elective semester (first semester of year 2) of the master’s programme it is possible to take electives from our other master’s programme or relevant master’s programmes at Maastricht University (maximum of 13 ECTS outside DKE) or to participate in a research project, a company internship or a study abroad semester at one of our partner universities. Please contact exchange officer and/or the Study Adviser for more information.

** The Research Project 1 will start in period 1.1 and 1.2 with weekly meetings. The credits for the project will become available at the end of period 1.3. The Research Project 2 will start in period 1.4 and 1.5 with weekly meetings. The credits for the project will become available at the end of period 1.6.

Part of the Research project is a Project skill programme (all elements are mandatory). Among the competences thought are advanced presentation and academic writing skills.
Period 1.1

Datamining (Code: KEN4113)
Examiner: Dr. E.N. Smirnov
Desired Prior Knowledge: Statistics and Basic Machine Learning
Prerequisites: None.
Course Description: Data mining is a major frontier field of computer science. It allows extracting useful and interesting patterns and knowledge from large data repositories such as databases and the Web. Data mining integrates techniques from the fields of databases, machine learning, statistics, and artificial intelligence. This course will present the state-of-the-art techniques of data mining. The lectures and labs will emphasize the practical use of the presented techniques and the problems of developing real data-mining applications. A step-by-step introduction to data-mining environments will enable the students to achieve specific skills, autonomy, and hands-on experience. A number of real data sets will be analysed and discussed.
Knowledge and understanding: Students will acquire knowledge on data preparation, data preprocessing, feature selection/generation, data mining, and model validation.
Applying knowledge and understanding: When confronted with real-life problems, students will be able to identify data-analysis tasks. Then, they will be able to apply data-mining techniques for supervised and unsupervised data-analysis. If necessary, students will be able to design data-mining algorithms specific for the tasks they have.
Making judgements: Students will be able to assess the quality of data-mining models, processes, results, and tools.
Communication: Students will be able to present the results of different stages of data-mining processes to specialists or non-specialists.
Learning skills: Students will be able to recognize their own lack of knowledge and understanding and take appropriate action such as consulting additional material or other sources of help.
Study material: Course notes, slides, and other information made available.
Recommended literature:
Exam: Written exam (80%) + practical assignments (20%).
ECTS: 6

Foundations of Agents (Code: KEN4115)
Examiner: Dr. ir. ing. N. Roos.
Description: Agents are autonomous computer programs, robots, humans, etc. Agents operate in some environment, which they can observe, and in which they can realize objectives through the execution of actions. Examples of environment in which agents can operate, are computer game environments, the internet, and also the physical world is case of robots and humans.
In this course, we address the problem of how an agent can act optimally in order to realize its objectives. We will answer this question by investigating how we can formally specify the agent’s environment, the agent’s objectives, the observations the agent can make and the actions it can execute. We use the formal model to investigate how the agent can determine an (optimal) behaviour realizing its objectives.
The following formal models will be investigated:
• Markov Decision Processes,
• Partially Observable Markov Decision Process,
• logic-based models such as Epistemic Logic, Doxastic Logic, Dynamic Logic, and BDI logics, and
• Game Theory.

Some examples of methods for determining the agents optimal behaviour addressed in the course are: Value and Policy Iteration, Q-Learning, Planning, etc.

Knowledge and understanding:
• The student will be able to explain formal models for describing agents.
• The student will be able to verify whether the formal models are correct.
• The student will be able to explain the underlying assumptions of each formal model.
• The student will be able to compare and discuss the differences between the formal models.
• The student will be able to analyse important properties of formal models for describing agent.

Applying knowledge and understanding:
• The student will be able to apply formal models for describing agents to solve practical problems.
• The student will be able to assess which formal model is most suited for addressing a practical problem.
• The student will be able to make a practical implementation of a formal model.

Making judgments:
• The student will be able to analyse which formal model of an agent is adequate for specific problem domains.
• The student will be able to select the formal model of an agent that is adequate for specific problem domains.

Communication:
• The student will be able to explain to its peers why a formal model of an agent is adequate for specific problem domains.
• The student will be able to explain to its peers how a formal model of an agent should be applied for a specific problem.

Learning skills:
• The student will be able to explain to its peers why a formal model of an agent is adequate for specific problem domains.
• The student will be able to explain to its peers how a formal model of an agent should be applied for a specific problem.

Study material: Syllabi, scientific papers.
Recommended literature: none.

Examination: Written exam at the end of the course. A bonus of 1.0 point can be earned by a series of bonus assignments.

ECTS: 6

Intelligent Search & Games (Code: KEN4123)
Examiners: Prof. dr. M.H.M. Winands and dr. C. Browne.
Desired Prior Knowledge: Data Structures & Algorithms
Description: In this course, the students learn how to apply advanced techniques in the framework of game-playing programs. Depending on the nature of the game, these techniques can be of a more or less algorithmic nature. The following subjects will be discussed:
(1) Basic search techniques. Alpha-beta; A*.
(2) Advanced search techniques. IDA*; B*, transposition tables; retrograde analysis
and endgame databases; proof-number search and variants; multi-player search methods; Expectimax and *-minimax variants.

(3) Heuristics. World representations; killer moves; history heuristic, PVS; windowing techniques; null-moves; forward-pruning techniques; selective search, GOAP.

(4) Monte Carlo methods. Monte Carlo tree search (MCTS) techniques, enhancements and applications; AlphaGo and AlphaZero approaches.

(5) Game design. Evolutionary game design; game quality metrics; self-play evaluation; procedural content generation (PCG); puzzle design.

Knowledge and understanding: The student can explain basic and advanced search techniques and can identify which of them to use either in a game context, or in problems with a similar structure.

Applying knowledge and understanding: Students have obtained the knowledge to develop, program, analyse, and apply advanced techniques autonomously to a wide variety of problems. They will also learn that adapting known techniques to fit a given problem can achieve a better performance.

Making judgements: Students will be able to judge the quality of approaches (systems or scientific publications) based on the techniques taught.

Communication: Students will be able to present the results of their game programs and search algorithms to specialists or non-specialists.

Learning skills: Students will be able to familiarize themselves with Game AI techniques beyond the scope of the course in order to solve a problem.

Study material: Course notes and other information made available.

Recommended literature:

Exam: Written exam (50%) + a large practical task (50%).

ECTS: 6

Mathematical Optimization (code: KEN4211)

Examiner: Dr. P. Collins

Tutor(s):

Desired Prior Knowledge: Simplex algorithm, Calculus, Linear Algebra.

Description: Optimization (or “Optimisation”) is the subject of finding the best or optimal solution to a problem from a set of potential or feasible solutions. Optimization problems are fundamental in all forms of decision-making, since one wishes to make the best decision in any context, and in the analysis of data, where one wishes to find the best model describing experimental data. This course treats two different areas of optimization: nonlinear optimization and combinatorial optimization. Nonlinear optimization deals with the situation that there is a continuum of available solutions. A best solution is then usually approximated with one of several available general-purpose algorithms, such as Brent’s method for one-dimensional problems, Newton, quasi-Newton and conjugate gradient methods for unconstrained problems, and Lagrangian methods, including active-set methods, sequential quadratic programming and interior-point methods for general constrained problems. Combinatorial optimization deals with situations that a best solution from a finite number of available solutions must be chosen. A variety of techniques, such as linear programming, branch and cut, Lagrange relaxation
dynamic programming and approximation algorithms are employed to tackle this type of problems. Throughout the course, we aim to provide a coherent framework for the subject, with a focus on consideration of optimality conditions (notably the Karush-Kuhn-Tucker conditions), Lagrange multipliers and duality, relaxation and approximate problems, and on convergence rates and computational complexity. The methods will be illustrated by in-class computer demonstrations, exercises illustrating the main concepts and algorithms, and modelling and computational work on case studies of practical interest, such as optimal control and network flow.

**Knowledge and understanding:** By the end of this course, students will have a strong foundation in nonlinear and combinatorial optimization. You will be able to formulate real-life problems as optimization problems. You will understand optimality conditions, including the Karush-Kuhn-Tucker conditions and be able to test for optimality. You will know how to solve a variety of general optimization problems, including constrained nonlinear problems, and (mixed-) integer linear problems. You will understand notions of duality and Lagrange multipliers, and be able to apply techniques based on relaxation and approximation.

**Applying knowledge and understanding:** Students will know the advantages and disadvantages of different methods, and be able to choose an appropriate method for a given problem. You will be able to implement and test optimization algorithms on a computer. You will be able to apply your knowledge to the solution of practical problems and in developing new efficient algorithms.

**Making judgements:** Students will be able to select an appropriate solution method for a given optimization problem, and judge the quality of the solution obtained.

**Communication:** Students will be able to discuss the development and use of optimization algorithms.

**Learning skills:** Students will learn how to develop and implement mathematical methods, select and evaluate algorithms, and formulate mathematical model of real-world problems.

**Study material:** Lecture notes, handouts.

**Recommended literature:**
1. Numerical Optimization, by Nocedal and Wright (Springer)
2a. Combinatorial Optimization, Algorithm and Complexity, by Papadimitriou and Steiglitz (Dover Publications), or

**Exam:** Written exam, closed book with formula sheet (100%)

**ECTS:** 6

**Signal and Image Processing (Code KEN4222)**

**Examiner:** Dr J. Karel and dr. P. Bonizzi

**Desired Prior Knowledge:** Linear algebra, Calculus, basic knowledge of Matlab. Some familiarity with linear systems theory and transforms (such as Fourier and Laplace) is helpful.

**Prerequisites:** None.

**Description:** This course offers the student a hands-on introduction into the area of digital signal and image processing. We start with the fundamental concepts and mathematical foundation. This includes a brief review of Fourier analysis, z-transforms and digital filters. Classical filtering from a linear systems perspective is discussed. Next wavelet transforms and principal component analysis are introduced. Wavelets are used to deal with morphological structures in signals. Principal component analysis is used to extract information from high-dimensional datasets.
We then discuss Hilbert-Huang Transform to perform detailed time-frequency analysis of signals. Attention is given to a variety of objectives, such as detection, noise removal, compression, prediction, reconstruction and feature extraction. We discuss a few cases from biomedical engineering, for instance involving ECG and EEG signals. The techniques are explained for both 1D and 2D (images) signal processing. The subject matter is clarified through exercises and examples involving various applications. In the practical classes, students will apply the techniques discussed in the lectures using the software package Matlab.

Knowledge and understanding: Students are able to explain fundamental concepts of signal and image processing and their mathematical foundation. They are able to 1) describe various types of filters and their properties, 2) explain orthogonal wavelet filter banks and describe their properties, 3) explain a construction scheme and elicit a wavelet-based noise-filtering scheme, 4) explain principal component analysis and empirical signal processing techniques and how they complement the other techniques discussed.

Applying knowledge and understanding: Students are able to use the various techniques discussed during the lectures to solve real-world problems, such as being able to apply wavelet filtering and principal component analysis on various signals. They are also able to analyse a signal by using Matlab, and independently interpret the outcome of an analysis.

Making judgements: Students are able to assess what technique is suited for a signal processing problem at hand, and to independently and critically look at a signal or image, and understand if and what type of pre-processing is required.

Communication: Students are able to communicate signal and image processing techniques and strategies.

Learning skills: Students are able to independently master signal and image processing techniques, from classical signal processing techniques to more empirical techniques.


Exam: Written exam/Computer exam.

ECTS: 6

Stochastic Decision Making (Code: KEN4221)

Examiner: Dr. G. Schoenmakers

Prerequisites: Probability & Statistics.

Course Description: Any realistic model of a real-world phenomenon must take into account the possibility of randomness. That is, more often than not, the quantities we are interested in will not be predictable in advance but, rather, will exhibit an inherent variation that should be taken into account by the model. Mathematically, this is usually accomplished by allowing the model to be probabilistic in nature. In this course, the following topics will be we discussed:

(1) Concepts of probability theory: Random variables, probability distribution functions, density functions, conditional probability, expectation and variance.
(2) Finding probabilities, expectations and variances of random variables in complex probabilistic experiments.
(3) Discrete and continuous time Markov chains and related stochastic processes like random walks, branching processes, Poisson processes, Birth and Death processes, and queueing theory.
(4) Markov decision problems.
Knowledge and understanding: In this course, the students acquire tools for modelling complex processes involving randomness, providing a basis for originality in developing and/or applying ideas in a research context.

Applying knowledge and understanding: When confronted with complex problems that involve probabilistic experiments, students have the tools to create and analyse appropriate models.

Making judgements: The students are able to analyse complex problems as stochastic processes and solve them. Furthermore, students can find optimal solutions in decision problems that are based on these stochastic processes.

Communication: The students will be able to communicate their conclusions and the underlying rationale to expert and non-expert audiences.

Learning skills: The students have obtained the skills to study related material in a largely autonomous manner.

Study material: Introduction to Probability Models by Sheldon M. Ross (9th or 10th ed.) + Lecture notes that are provided via Student Portal.

Recommended literature: Probability: A Lively Introduction by Henk Tijms.

Exam: Written exam.

ECTS: 6

Period 1.2

Advanced Concepts in Machine Learning (Code: KEN4154)

Examiner: Dr. ir. K. Driessens

Desired Prior Knowledge: Machine Learning

Prerequisites: None.

Description: This course will introduce a number of advanced concepts in the field of machine learning such as Support Vector Machines, Gaussian Processes, Recommender Systems, Deep Neural Networks, Reinforcement learning, etc. All of these are approached from the view that the right representation is imperative for machine learning solutions. Additionally, different knowledge representation formats used in machine learning are introduced. This course counts on the fact that basics of machine learning were introduced in other courses so that it can focus on more recent developments and state of the art in machine learning research. Labs and assignments will give the students the opportunity to implement or work with these techniques and will require them to read and understand published scientific papers from recent Machine Learning conferences.

Knowledge and understanding: Students can explain, construct and adapt powerful machine learning techniques, most with a statistical background. Students recognise the need for non-standard techniques and representations that can be used for complex/structured data. They can explain the strengths and weaknesses of different machine learning approaches.

Applying knowledge and understanding: Students will be able to select, adapt and apply a number of advanced machine learning approaches. They will be able to select the correct representation for a machine-learning problem and to translate a machine-learning problem into a suited representational format.

Making judgements: Students will be able to judge which machine learning approach and data-representation is best suited. They will also be able to comprehend and judge machine-learning research.

Communication: Students will be able to relate different machine learning techniques to each other and explain their working, benefits and disadvantages to non-experts. They will also be able to discuss the need and use of structured representation with both experts and non-experts.
Learning skills: Students will be able to relate information from different sources, and read process and evaluate recent research developments in the field of machine learning.

Study material: Slides and collected notes and chapters from freely available books and course notes.

Recommended literature: Pattern Recognition and Machine Learning - C.M. Bishop; Bayesian Reasoning and Machine Learning - D. Barber; Gaussian Processes for Machine Learning - C.E. Rasmussen & C. Williams; The Elements of Statistical Learning - T. Hastie et al.

Exam: Students are graded using a number of assignments (20%), a report and presentation on a recent scientific paper or a machine learning application (20%) and a written exam (60%).

ECTS: 6

Applications of Image and Video Processing (Code: KEN4244)

Examining: Dr. A. Briassouli


Prerequisites: None.

Description: Applications of image and video processing will be presented, and connections to basic algorithms will be demonstrated. We will examine some of the most popular and widespread applications, namely security, surveillance, medical, traffic monitoring, astronomy, farming, culture. The methods used in these applications will be analysed in class and common characteristics between them will be explained. Students will be able to suggest further applications of interest to them and bring relevant literature to the class.

Knowledge and understanding: Students will acquire a wide-ranging understanding of the latest trends in image and video processing methods and how these are applied in real world applications. They will obtain insights on common problems encountered in these applications, and how they can be tackled through advanced image and video processing algorithms.

Applying knowledge and understanding: The knowledge and understanding obtained in this class will be demonstrated in mini projects based on State of the Art research.

Making judgements: Through the presentation of various applications of image and video processing, students will be able to analyse problems in the real world, and understand how to best address them.

Communication: Part of the class will include homework where students will carry out a short literature review and implementation of mini projects on applications that interest them. They will be taught how to communicate them succinctly and effectively, maintaining a balance between overall understanding and technical depth.

Learning skills: Students will obtain a spherical comprehension of connections between machine learning and image/video/signal processing, as well as their practical in a wide range of applications in our daily life. They will be able to identify the methods needed in different applications of image and video processing, propose a plan for solving the corresponding problems, and justify it.

Study material: Lecture slides, selected papers.

- A Bovik, Handbook of Image and Video processing

Recommended literature:

Exam: Mini projects (50%) and final exam (50%).

ECTS: 6
Model Identification and Data Fitting (Code: KEN4242)


Tutor(s): None.

Desired Prior Knowledge: Basic knowledge of Matlab and some familiarity with linear systems theory and transforms (such as Fourier and Laplace) is helpful. This course offers a useful prior knowledge for the course Symbolic Computation and Control.

Prerequisites: Linear Algebra, Mathematical Modelling, Probability and Statistics.

Description: This course is devoted to the various practical and theoretical aspects which involve the estimation (the identification) of a mathematical model within a given model class, starting from a record of observed measurement data (input-output data). First, we address distance measures, norms, and criterion functions. Then we discuss the prediction error identification of linear regression models, with special emphasis on the various interpretations of such models (deterministic, stochastic with Gaussian white noise and maximum likelihood estimation, stochastic in a Bayesian estimation context) and on numerical implementation aspects (recursion, numerical complexity, numerical conditioning and square root filtering). Next, we study identification within the important class of auto-regressive dynamical models, to which the Levinson algorithm applies. Other related topics receiving attention are identifiability, model reduction and model approximation. Some techniques for the estimation of linear dynamical i/o-systems are illustrated with the system identification toolbox in Matlab.

Knowledge and understanding: Students learn to recognize the various aspects that play a key role in building a mathematical model from measurement data: the choice of model class (and order), the choice of parameterization, the criterion of fit, the model estimation method, the quality of the measurement data, and the validity of the estimated model.

Applying knowledge and understanding: Students are able to 1) estimate models from measurement data, particularly linear regression models and auto-regressive models, 2) to assess the quality of a (linear regression) model, and 3) assess whether a model is identifiable.

Making judgements: Students are able to predict and judge the quality of models that can be obtained from a record of measurement data.

Communication: Students have obtained the skills to motivate the choice of a model class, the model order and an estimation method to identify a model from measurement data, to interpret the identification outcomes and to explain all this to specialists and non-specialists.

Learning skills: Students are able to read and interpret scientific literature on model estimation and system identification, and to use Matlab and work out ideas computationally.

Study material: Syllabus, provided electronically through the “My Courses” section of the Student Portal.

Recommended literature:

Exam: Written exam.

ECTS: 6
**Multi-Agent Systems (Code KEN4111)**  
*Examiner:* Prof. dr. G. Weiss.  
*Desired Prior Knowledge:* Introduction to Computer Science 1 and 2.  
*Description:* Multi-agent systems are systems composed of multiple interacting intelligent agents, where an agent is a computational entity such as a software program or a robot that is situated in some environment and that to some extent is able to act autonomously in order to achieve its design objectives. The field of multi-agent systems has its origin in the late 1970s and today is an established and vibrant topic in computer science. Multi-agent systems are an enabling technology for applications that rely on distributed and parallel processing of data, information and knowledge in complex – networked, open and large-scale – computing environments. With advancing technological progress in inter-connectivity and interoperability of computers and software such applications are becoming standard in a variety of domains such as e-commerce, logistics, supply chain management, telecommunication, health care, and manufacturing. The course covers important conceptual, theoretical and practical foundations of multi-agent systems. Examples of topics treated in the course are agent-agent communication, automated negotiation and argumentation in cooperative and competitive settings, multi-agent learning and planning, automated decision making based on mechanisms such as voting and auctioning, and development and engineering of agent-based systems. In the practical part of the course students have the opportunity to apply the learnt multi-agent concepts, algorithms and methods.  
*Knowledge and understanding:* The student is able to describe existing and design novel agent-agent coordination mechanisms and interaction principles, and can explain and analyse their strengths and shortcomings.  
*Applying knowledge and understanding:* The student is be able to apply the gained knowledge in concrete application scenarios and practical applications.  
*Making judgements:* The student is be able to judge for a given problem whether and in how far it is beneficial to use a multi-agent approach for its solution.  
*Communication:* The student is able to motivate and explain benefits and shortcomings of their usage in a given application, and thereby showing sufficient understanding of multi-agent concepts.  
*Learning skills:* The student is able to study autonomously multi-agent system literature, including, in particular, literature describing new developments in the methods and techniques covered in this course  
*Study material:* Course slides; supplementary material to be announced.  
*Recommended literature:*  
*Examination:* Practical and reading assignments (30%) and written exam (70%)  
*ECTS:* 6

**Information Security (KEN4245) *with reservation**  
*Examiner:* Dr. A. Zarras  
*Tutor:* Dr. A. Zarras  
*Desired Prior Knowledge:* Introduction to Computer Science (1 & 2), Data Structures and Algorithms, Software Engineering, Databases  
*Description:* Information security is a specialized field in computer security that is
designed to detect and prevent threats against digital information. As the Internet evolves and computer networks become bigger and bigger, a data breach can have a severe impact to an organization as sensitive and confidential data can be accessed and disclosed in an unauthorized fashion. This illicit access to the data may involve bad publicity and financial losses for the organization. This course will cover a wide range of topics in information security, trying to strike balance between core concepts and recent advancements. The main objectives of the course are to instil into the students a heightened awareness of the need to protect data and resources from disclosure as well as to guarantee the authenticity of data and messages. Some of the topics we will explore include information assurance, access control, cryptography, and authentication.

**Knowledge and understanding:** Students are able to understand the problems of information security. In detail, they will understand what the threats they may face as data scientists and how to effectively tackle them.

**Applying knowledge and understanding:** Students will be able to demonstrate their knowledge through graded assignments.

**Making judgements:** By understanding the fundamentals of information security and by realizing their assignments, the students will be able to understand and avoid common mistakes when operating with large and sensitive datasets.

**Communication:** The class will consist of lectures in which several information security issues will be discussed. In parallel, there will be bonus assignments where the students will have to solve some of the most important issues we discussed in classroom. This way they will be able learn all the described principles in depth.

**Learning skills:** Students will be able to protect the integrity and authenticity of their data as well as design and develop tools that enforce information security.

**Exam:** Assignments and Project

**ECTS:** 6

**Period 1.3**

**Research Project AI 1 & DSDM 1 (Code KEN4131 & Code KEN4231)**

**Coordinator:** Dr. ir. K. Driessens

**Tutors:** Dr. ir. K. Driessens & Dr. G. Schoenmakers

**Desired Prior Knowledge:** None.

**Prerequisites:** None.

**Description:** The research project takes place during the three periods of the semester. Project topics are presented at the start of the semester and assigned to students based on their preferences and availability. The emphasis in the first phase is on initial study of the context set out for the project and the development of a project plan. In the second period, the goal is to start modelling, prototyping and developing. In period 3, the implementation, model and/or experiments set out in the project plan has to be finished and reported on. At the end of period 1 and 2, a progress presentation takes place. The project results in a project presentation, a project report and possibly a public website and/or product.

**Knowledge and understanding:** Students get to know and possibly contribute to state of the art methods within the fields of Artificial Intelligence and/or Data Science for Decision Making to answer an open question.

**Applying knowledge and understanding:** Student write their own research plan in coordination with a staff member (plus possibly outsiders) who act as clients with an open question. Students with different backgrounds and from both masters work together in teams to build and evaluate an answer to an open question. Students
find, judge the suitability, apply, and evaluate state of the art techniques to answer questions and construct applications in the field of Artificial Intelligence and Data Science. Students apply the accumulated knowledge from other educational activities in application specific areas.

**Making judgements:** Students judge feasibility of tasks, attainability of goals, and the amount of work involved. Students think about the possible consequences of their work. Students evaluate state of the art and the applicability and scope of research results.

**Communication:** Students will learn to:
1. orally communicate and cooperate with peers
2. orally report on progress and intermediate results to superiors
3. orally negotiate and communicate with clients
4. communicate their ideas in written form, both for an academic and a general audience
5. give effective presentations

**Learning skills:** Students increase their own level of knowledge in a specialised sub-discipline of the field of Artificial Intelligence and/or Data Science. Students perform research into recent state of the art techniques. Students learn that the field of Artificial Intelligence and Data Science are constantly evolving beyond what is taught in class.

**Study material:** Slides provided at the end of joint information sessions. Literature provided by the project supervisors.


**Exam:** Phase 1: project plan + presentation (15%); Phase 2: layman’s website + presentation (15%); Phase 3: Project report + presentation (70%)

**ECTS:** 6

### Period 1.4

**Algorithms for Big Data (Code: KEN4254)**

**Examiner:** Dr. M. Mihalák

**Desired Prior Knowledge:** Discrete mathematics, algorithm design and analysis, elementary discrete probability

**Prerequisites:** None

**Description:** The emergence of very large datasets poses new challenges for the algorithm designer. For example, the data may not fit into the main memory anymore, and caching from a hard-drive becomes a new bottleneck that needs to be addressed. Similarly, algorithms with larger than linear running time take simply too long on very large datasets. Moreover, simple sensory devices can observe large amount of data over time, but cannot store all the observed information due to insufficient storage, and an immediate decision of what to store and compute needs to be made. Classical algorithmic techniques do not address these challenges, and a new algorithmic toolkit needs to be developed. In this course, we will look at a number of algorithmic responses to these problems, such as: algorithms with (sub-)linear running times, algorithms where the data arrive as a stream, computational models where memory is organized hierarchically (with larger storage units, such as hard-drives, being slower to access than smaller, faster storage such as CPU cache memory). New programming paradigms and models such as MapReduce/Hadoop will be discussed. We will also look at a number of topics from classical algorithm design that have undiminished relevance in the era of big data such as approximation algorithms and multivariate algorithmic analysis.
Knowledge and understanding: Students will know, exemplified on selected topics, what can be provably achieved when designing and analysing algorithms for very large datasets, and will know some of the most successful state-of-the-art algorithmic techniques for dealing with algorithmic challenges posed by large data sets.

Applying knowledge and understanding: Students will be able to adjust and apply the gained knowledge about algorithmic techniques to various algorithmic challenges of handling large datasets.

Making judgements: Students will be able to categorize large-scale problems according to their computational feasibility, and select the appropriate algorithmic response.

Communication: Students will be able to reason about computational problems and algorithms addressing the problems in a clear, exact, and unambiguous way.

Learning skills: Additionally to the guiding material provided by the lecture, the students will autonomously search, read, and study the details from various sources.

Study material: Will be provided throughout the lecture.

Recommended literature: None.

Exam: Written exam (80%) at the end of the course and graded exercises (20%) throughout the course.

ECTS: 6

Building and Mining Knowledge Graphs (Code: KEN4256)

Examiner: Prof. dr. M. Dumontier and Dr. K. Moodley

Tutor(s): None.

Desired Prior Knowledge: Introduction to Computer Science

Prerequisites: None.

Description: Knowledge graphs are large-scale, machine-processable representations of entities, their attributes, and their relationships. Knowledge graphs enable both people and machines to explore, understand, and reuse information in a wide variety of applications such as answering questions, finding relevant content, understanding social structures, and making scientific discoveries. However, the sheer size and complexity of these graphs present a formidable challenge particularly when mining across different topic areas.

In this course, we will examine approaches to construct and use knowledge graphs across a diverse set of applications using cutting-edge technologies such as machine learning and deep learning, graph databases, ontologies and automated reasoning, and other relevant techniques in the area of data mining and knowledge representation.

Knowledge and understanding: Students will be able to describe:
- The nature and attributes of a Knowledge Graph
- Examples of Knowledge Graphs;
- Representations for Knowledge Graphs
- Applications of Knowledge Graphs
- Advantages and disadvantages of Knowledge Graphs as compared to other formalisms
- Approaches and challenges in constructing and maintaining Knowledge Graphs
- Approaches and challenges in finding, using and mining Knowledge graphs
- What FAIR is and how it relates to Knowledge Graphs

Ethical, legal & social issues around Knowledge Graphs

Applying knowledge and understanding: Students will be able to identify requirements and steps to convert knowledge in traditional data formats to Knowledge Graph formats. Students will also be able to implement such strategies. Students will be able to query Knowledge Graphs (for instance using SPARQL query
Students study in small groups, and the knowledge that is gained is applied to projects linked to the curriculum. The results of these projects are presented to teachers and fellow students.
language) to answer basic to intermediately advanced questions. Students will be able to implement basic reasoning strategies on Knowledge Graphs to answer intermediately advanced questions, which cannot be answered by SPARQL queries alone. Students will be able to implement popular methods to integrate different data sources by transferring them into a Knowledge Graph. Students will be able to enrich existing Knowledge Graphs with missing information using basic predictive algorithms. Students will be able to perform basic data quality assessment on Knowledge Graphs. Students will be able to assess the degree of compliance that Knowledge Graphs have with FAIR principles.

**Making judgements:** Students will be able to select which tools are most suitable for constructing, querying, visualising & reasoning with Knowledge Graphs. Students will be able to differentiate between different types of Knowledge Graphs, according to their representation, coverage and content. Students will be able to select which Knowledge Graph is appropriate for answering a particular question. Students will be able to diagnose incompleteness in a Knowledge Graph with respect to answering a particular question. Students will be able to evaluate the data quality and FAIRness of a Knowledge Graph.

**Communication:** Students will be able to explain the advantages of representing information on the web in Knowledge Graphs. Students will be able to communicate the steps required to convert information to a Knowledge Graph format. Students will be able to communicate to non-experts the main content and representational components of a Knowledge Graph. Students will be able to outline to non-experts the steps required to answer a question by querying a Knowledge Graph.

**Learning skills:** Students will be able to reflect critically on the challenges and open problems remaining in Knowledge Graphs research. Students will be able to formulate and propose strategies to answer complex questions using Knowledge Graphs. Students will be able to assess the feasibility of different combinations of methods for answering questions using Knowledge Graphs.

**Study material:** Material will be provided during the course in the form of handouts.


**Exam:** Individual project for application of knowledge and two assignments to demonstrate understanding of core concepts.

**ECTS:** 6

**Computational Statistics (Code KEN4258)**

**Examiner:** Dr. C. Seiler

**Desired Prior Knowledge:** Probability and Statistics

**Prerequisites:** None

**Description:** Complex and high dimensional data are abundant in academia and industry. At the same time, computers are cheap and powerful. These developments enable us to fit increasingly complex statistical models using computer intensive methods. We will model and analyze both independent and dependent data from real world problems. The course is hands-on; we will use Stan (platform for statistical modelling and high-performance statistical computation) and R (statistical programming language). Key topics: Statistical modelling, uncertainty quantification, Markov chain Monte Carlo, bootstrap resampling, permutation tests, and causal inference.

**Knowledge and understanding:** Knowing a wide range of modern statistical models and computational tools to draw inferences will provide the foundations for analyzing complex data in academia and industry.

**Applying knowledge and understanding:** Students will be able to:
1. Tidy raw data obtained from large databases
2. Build complex statistical models
3. Choose computational tools to perform inference
4. Create reproducible analysis workflows to communicate results

**Making judgements:** In this course, we will discuss one of the most important aspects of analyzing data: being skeptical of results and avoiding wishful thinking. This will be accomplished by careful model checking and interpretation of the results.

**Communication:** Students will present their results using literate programming and reproducible workflows.

**Learning skills:** Students will be able to understand, apply, and extend papers from computational statistics journals.

**Study material:**

**Recommended literature:**
Selected chapters:
- Efron and Tibshirani (1993), An Introduction to the Bootstrap
- Hoff (2009), A First Course in Bayesian Statistical Methods
- Grolemund and Wickham (2017), R for Data Science
- Hernán and Robins (2019, forthcoming), Causal Inference

**Exam:** Written exam and assignments (focus on programming exercises, only few selected mathematical exercises).

**ECTS:** 6

**Autonomous Robotics Systems (Code: KEN4114)**

**Examiner:** Dr. R. Möckel.


**Prerequisites:** None

**Description:** Operating autonomously in unknown and dynamically changing environments is a core challenge that all robotic systems must solve to work successfully in industrial, public, and private areas. Currently popular systems that must demonstrate such capabilities include self-driving cars, autonomously operating drones, and personal robotic assistants. In this course, students obtain deep knowledge in creating autonomous robotic systems that can operate in and manipulate unknown and dynamically changing environments by autonomously planning, analysing, mapping, and modelling of such environments. Students learn to approach these challenging tasks through three main techniques: swarm intelligence, model-based probabilistic frameworks, and (mostly) model-free techniques from artificial evolution and machine learning.

**Knowledge and understanding:** Students gain a deep understanding of the challenges in autonomous robotic systems and how these challenges are addressed in state-of-the-art systems. Students learn about and practise techniques for autonomous mapping, localization, navigation, sensing, modelling robot motion, planning, and decision-making. Through the course, students obtain in-depth knowledge and hands-on experience in a variety of algorithms and techniques from machine learning, agent technology, and search techniques including Bayesian filters (like Kalman Filters, Extended Kalman Filters, Histogram Filters, and Particle Filters), artificial neural networks, evolutionary algorithms, and swarm intelligence.

**Applying knowledge and understanding:** After successful completion of the course, students will have obtained in-depth knowledge to understand, adapt, apply, and combine a number of advanced control approaches to create the next generation of...
autonomous robotics systems. Students obtain the ability to select from a variety of available tools feasible solutions for the complex and rather ill defined problem domains of autonomous robotic systems and to predict the resulting consequences of their choices. Furthermore, students learn how to choose, apply, formulate, and validate models of autonomous robotic systems and of appropriate control techniques from artificial intelligence for these systems.

**Making judgements:** Students will be able to comprehend and to critically judge scientific publications on autonomous systems, artificial evolution, and swarm intelligence. From this literature, students are able to search for and to critically process information to solve given ill-defined but in practise highly relevant problems in autonomous systems. Students are able to critically discuss social, economic, and ethical consequences of artificial intelligence and autonomous decision-making.

**Communication:** Students learn to critically discuss challenges and professional solutions in autonomous robotic applications with both experts and non-experts.

**Learning skills:** The course prepares students to work on robotic applications in professional research and business environments. Students will be able to autonomously acquire new skills and knowledge to develop, program, analyse, and apply advanced techniques to a wide variety of problems.


Lecture material and publications provided during the lecture.

**Recommended literature:**

**Examination:** The final course grade is 80% of the final written “closed-book” exam grade plus 20% of the practical group assignments grades

**ECTS:** 6

**Dynamic Game Theory (Code KEN4251)**

**Examiner:** Prof. Dr. F. Thuijsman and Dr. K. Stankova.

**Desired Prior Knowledge:** Students are expected to be familiar with basic concepts from linear algebra, calculus, Markov chains and differential equations.

**Prerequisites:** None

**Description:** The course will focus on non-cooperative games and on dynamic games in the following order: matrix and bimatrix games, repeated games, Stackelberg games, differential games, specific models of stochastic games, evolutionary games. These are games in which the players are acting as strategic decision makers, who cannot make binding agreements to achieve their goals. Instead, threats may be applied to establish stable outcomes. Besides, relations with population dynamics and with “learning” will be examined. Several examples will be taken from biological settings. Knowledge and understanding Students are able to recognize and classify the main types of dynamic games, i.e. repeated games, stochastic games, Stackelberg games, differential games, and evolutionary games and formulate the main solution concepts value, optimal strategies, Nash- and Stackelberg equilibrium

**Applying knowledge and understanding:** Students are able calculate solutions of the different types of dynamic games

**Making judgements:** Students are able to explain advantages and disadvantages of different solution concepts. They are able to judge correctness of solutions presented

**Communication:** Students are able to explain and defend correctness of their
solutions

**Learning skills:** By the end of the course, students will be able to autonomously and critically reflect upon the pros and cons of different types of games for modelling competition and cooperation. This includes considerations on the computational aspects with respect to different solution concepts.

**Study material:** Handouts will be provided.

**Recommended literature:** none.

**Exam:** There will be a closed book written exam at the end of the course.

**ECTS:** 6

**Advanced Natural Language Processing (Code: KEN4259)**

**Examiner:** Dr. J. Niehues

**Desired Prior Knowledge:** Advanced Concepts in Machine Learning

**Prerequisites:** none

**Description:** How do I say, “Where is the next Italian restaurant” in Dutch? Can I get a summary of today’s lecture? When were artificial neural networks developed? Computers able to answer these questions are a long-time dream of humankind and currently, we see first programs to solve these problems. This course will provide the skills and knowledge to develop state-of-the-art (SOTA) solutions for these natural language processing (NLP) tasks.

After a short introduction to traditional statistical approaches to NLP, the course will focus on deep learning techniques to solve these problems. In the first part of the course, we will investigate methods to model sequence labeling tasks like Named Entity recognition or Part-of-speech techniques. The second part of the lecture will focus on sequence-to-sequence models, a very powerful model to solve many NLP tasks like machine translation, summarization and question answering.

In this course, major challenges when building the systems will be address: representing words in neural networks, neural network architectures to model language, methods to train complex models and algorithms to find the most probable output.

**Knowledge and understanding:** Student will be taught state-of-the-art deep learning techniques for natural language processing, especially sequence labeling and sequence-to-sequence models. They will learn techniques to address the major challenges when building a natural language processing tool.

**Applying Knowledge and understanding:** The achievements in deep learning have significantly improved the quality of state-of-the-art methods for natural language processing. With the knowledge acquired in the course, students will be able to build SOTA solutions.

**Making judgements:** Students will be able to analyze the specific challenges of a task in NLP. Based on the gather knowledge on different ways to model tasks they are able to select and implement a fitting model to solve the task.

**Communication:** Through small research projects, students will be enabled to communicate their findings and explain the rationale behind their choices in deep learning techniques for natural language processing.

**Learning skills:** After successful completion of the course, students will be able to develop natural language processing tools and perform research on new ideas in the field.

**Study material:** Mostly based on the lecture notes and the provided material including recent papers published in this field.

**Recommended literature:** Papers published in top international conferences and journals in machine learning field.

**Exam:** Group-assignment (30%), final exam (70%)

**ECTS:** 6
Period 1.5

**Computer Vision (Code: KEN4255)**

**Examiner:** Dr. M. Popa and Dr. S. Asteriadis

**Desired Prior Knowledge:** Basic knowledge of Matlab, linear algebra and machine learning. This course offers the basics on image processing although prior knowledge is also a plus.

**Description:** Can we make machines look, understand and interpret the world around them? Can we make cars that can autonomously navigate in the world, robots that can recognize and grasp objects and, ultimately, recognize humans and communicate with them? How do search engines index and retrieve billions of images? This course will provide the knowledge and skills that are fundamental to core vision tasks of one of the fastest growing fields in academia and industry: visual computing. Topics include introduction to fundamental problems of computer vision, mathematical models and computational methodologies for their solution, implementation of real-life applications and experimentation with various techniques in the field of scene analysis and understanding. In particular, after a recap of basic image analysis tools (enhancement, restoration, color spaces, edge detection), students will learn about feature detectors and trackers, fitting, image geometric transformation and mosaicing techniques, texture analysis and classification using unsupervised techniques, object classification and face recognition, camera models, epipolar geometry and 3D reconstruction from 2D views.

**Knowledge and understanding:** Students will be able to apply the most suitable techniques for image pre-processing (e.g. enhancement, restoration), feature extraction, texture analysis, perspective geometry, camera models and topics on object recognition. Students will be able to identify the most suitable techniques in a series of visual computing problems.

**Applying knowledge and understanding:** The students will be able to choose and/or construct solutions in a variety of professional/vocational contexts requiring image processing and computer vision (robotics, manufacturing, AI, web applications, surveillance). They will be able to build and assess methodologies for handling real-world complex problems in computer vision, making use of pre-existing data for training their models.

**Making judgements:** The students will be able to choose and combine the right methods to tackle real-world computer vision problems, captured in real-life settings and having no obvious solutions. They will be able to propose and build techniques combining computer vision methods along with machine learning instruments for scene understanding and object recognition.

**Communication:** Through small research projects, students will be able to communicate their findings and explain the rationale behind their choices in computer vision techniques for image/video analysis.

**Learning skills:** After successful completion of the course, students will be able to analyse images and videos and retrieve or process content in order to derive useful information, applicable in a variety of domains (e.g. satellite imagery, surveillance, robotics, medical imaging).

**Study material:**
- Lecture slides and provided notes

**Recommended literature:** Digital Image Processing”, Rafael C. Gonzalez & Richard E.
Information Retrieval and Text Mining (Code: KEN4153)

Examiner: Prof dr. ir J.C. Scholtes
Tutor(s): None.
Desired Prior Knowledge: None.
Prerequisites: None.
Description: Text mining refers generally to the process of extracting interesting and non-trivial information and knowledge from unstructured text. Text mining encompasses several computer science disciplines with a strong orientation towards artificial intelligence in general, including but not limited to information retrieval (building a search engine), statistical pattern recognition, natural language processing, information extraction and different methods of machine learning, clustering and ultimately data visualization. An important difference with standard information retrieval (search) techniques is that they require a user to know what he or she is looking for, while text mining attempts to discover information in a pattern that is not known beforehand. This is very relevant, for example, in criminal investigations, legal discovery, (business) intelligence, sentiment- & emotion mining or clinical research.

Knowledge and understanding: The student will be able to select, understand and apply different phases and methods used to create successful Information Retrieval and Text Mining applications. In addition, the student learns to evaluate the quality of such methods according to best-practice standards as used in the field.

Applying knowledge and understanding: Students will be able to recognize applications of text mining and information retrieval in different domains such as legal services, medical research, regulatory oversight, compliance, humanities, and customer services. After the course, the student can formulate an opinion or course of action when dealing with text-based KE-problems based on incomplete, limited and in part unreliable information. After the course, students can apply their knowledge and understanding in a manner that shows a scientific approach to their work or vocation. They are able to handle complex and ill-defined text-based problems for which it is not a priori known if there is an appropriate solution, they know how to acquire the necessary information to solve the sub-problems involved, and they know how to proceed with problems for which there is no standard or reliable route to the solution.

Making judgements: Upon completion of the course, students are able to recommend the most appropriate methods from the fields of text mining and information retrieval when confronted with KE-problems involving textual and other forms of unstructured data.

Communication: Students are able to communicate the (dis)advantages of several methods from the field of text mining and information retrieval to both an audience of non-experts.

Learning skills: After the course, the student has developed those learning skills that are necessary for a successful further career in text mining or information retrieval at the highest professional level. The student will be able to continue to develop their text-mining and information retrieval skills. The student is able to detect missing knowledge and abilities and to deal with them appropriately by finding and consulting resources that can help them to fill the gaps and new developments.

Study material: A syllabus and copies of the course slides will be used.

**Exam:** The result of the practical exercises contributes 30% to the final examination of the course. The other 70% is determined by the theoretical exam. The theoretical exam is open book. For the practical exercise, students can select a research topic and a text corpus from the provided list (or another relevant open source collection) and implement a number of relevant text-mining operations by using open source text-mining tools. A number of relevant pre-processing operations, text mining operations, and visualizations have to be implemented. Proposals of work will have to be within one week after the start of the course, after which they will be reviewed. After approval, the students can start the implementation of their proposals. At the end of the course, each student or group shall write a report on the research and the results and the results shall be presented to the rest of the class.

**ECTS:** 6

**Deep Learning (Code KEN4257)**

**Examiner:** Dr. S. Mehrkanoon

**Desired Prior Knowledge:** Machine Learning

**Prerequisite:** Advanced Concepts of Machine Learning

**Description:** Conventional machine learning techniques were limited in processing data in their raw forms and many domain experts were required in transforming raw data into meaningful features or representations. Deep Learning techniques have revolutionized many application domains ranging from auditory to vision signal processing. In this course, we will study various concepts in deep architectures using both artificial neural networks as well as kernel-based models. Several deep learning models such as convolutional neural networks, auto-encoders, generative adversarial networks and their variants among other state-of-the-art models will be covered in depth. We will further study different types of deep architectures used for domain adaptation problems where one is encountered with heterogeneous datasets as well as multi-modal datasets. The regularization and optimization methods used in deep learning framework will be discussed. Introduction to open-source deep learning platforms will be given. This course will be equipped with a practical component, and students are expected to write their own deep learning code and test its performance on various problems. In addition they are strongly encouraged to participate in mini-projects (in a group or individual) targeting a conference paper.

**Knowledge and understanding:** By the end of this course, students will be able to explain aspects related to learning the model parameters, overfitting, model architecture, relation between neurons, layers among others. In addition, they will be able to implement the methodologies using deep learning libraries. The learned deep machine learning models combined with mathematics and optimization behind the models will enable students to gain more insight on capabilities of these models.

**Applying knowledge and understanding:** Students will be able to apply the above-mentioned deep learning skills, which they have acquired in their domain of interest. Explore and select new research directions and discover new things by driving new models.

**Making judgements:** Students will be able to implement and apply deep learning models for analyzing datasets in different tasks. They will be able to determine the best model for the given task.
Communication: Students will be able to discuss deep learning models and their results with scientists, engineers and both expert and non-expert.

Learning skills: By the end of the course, students will be able to autonomously follow up the recent trends in deep learning which are beyond the scope of the provided course materials.

Study material: Course notes and other information made available.

Recommended literature:
- Research Papers Published in high ranked journals and conferences.

Exam: Grades of the assignments (20%) + Grade for paper review, presentation (20%) + Grade for the exam (60%). Bonus: If students can come up with novel approaches to solve the given assignments and after consulting the results with me they can write a manuscript describing their findings and submit it to a conference for review can get additional 10% bonus on the assignments which could help them for the final grades. This bonus does not depend on the acceptance of the manuscript by the conference.

ECTS: 6

Planning and Scheduling (Code KEN4253)

Examiner: Dr. M. Mihalák

Desired Prior Knowledge: Data Structures & Algorithms. Discrete Mathematics. Graph Theory

Prerequisites: None

Description: In many real-world processes, particularly in industrial processes and logistics, decisions need to be taken about the time of the completion of (sub)tasks, and the decision about what production machines complete the tasks. There are often constraints on the order in which tasks, or ‘jobs’ can be performed, and there are usually capacity constraints of the machines. This leads to natural, industrially critical optimization problems. For example, a company might choose to buy many machines to process jobs, but then there is a risk that the machines will be underused, which is economically inefficient. On the other hand, too few machines, or an inappropriate ordering of tasks, may lead to machines spending a significant amount of time standing idle, waiting for the output of other machines, which are overcrowded with tasks. In this course, we look at various mathematical models and techniques for optimizing planning and scheduling problems, subject to different optimality criteria. We will discuss, among others, single-machine models, parallel-machine models, job-shop models, and algorithms for planning and scheduling (exact, approximate, heuristic) and we also touch upon the computational complexity (distinguishing between ‘easy’ and ‘difficult’ problems) of the underlying problems. Last but not least, we will also introduce integer linear programming as a uniform and generic tool to model and solve planning and scheduling problems.

Knowledge and understanding: Students will possess the mathematical and algorithmic tools to model and solve planning/scheduling problems. Students will be able to recognize real-world problems in the unified theory and established language of planning and scheduling.

Applying knowledge and understanding: Students will be able to apply the new techniques to various problems arising in real-world applications. Students will be able to deploy the standard algorithmic techniques, and be able to design new algorithmic solutions, and to argue about their performance properties.

Making judgements: Students will understand under which circumstances different planning/scheduling problems are computationally tractable, and will judge
algorithmic technique can be used to exactly or approximately solve these problems. **Communication:** Students will be able to analytically argue about correctness of the used algorithmic approaches. Students will be able to explain modelling approaches to planning and scheduling problems in the language of the theory of planning and scheduling.

**Learning skills:** Students will enhance their study skills such as time management, effective reading, critical thinking and reading, exact and unambiguous writing and formulating of ideas and statements, and reflection on marked work. Along the way, students will improve general learning skills such as self-motivation, careful listening and giving instructions, and openness to new knowledge. Students will also be exposed to autonomous self-study.

**Study material:** Appropriate study material will be provided throughout the course.

**Recommended literature:** None

**Exam:** Written exam (80%) at the end of the course, and graded exercises (20%) throughout the course.

**ECTS:** 6

**Symbolic Computation and Control (Code: KEN4252)**

**Examiner:** Prof dr. ir. R.L.M. Peeters.

**Desired Prior Knowledge:** Linear Algebra, Calculus, Mathematical Modelling.

**Course Description:** This course consists of two interrelated parts. In the first part, we focus on basic techniques for the digital control of linear dynamical systems using feedback. We start by addressing system stability and we discuss the technique of pole placement by state feedback to solve the regulation problem. Then we introduce state observers to solve the regulation problem by output feedback. Next, we extend our scope to tracking problems. This involves the design of additional dynamics to characterize the relevant class of reference signals, which are then integrated with the earlier set-up for output feedback. Finally, we discuss the classical topic of optimal control, which can be employed to avoid using prototype systems for pole placement, and which allows the user to design a feedback law by trading off the cost involved in generating large inputs against the achieved tracking accuracy. In the second part, we address computational issues, related to the field of systems and control. Classically, computers have been designed primarily to perform approximate numerical arithmetic. Modern software packages for mathematical computation, such as Maple and Mathematica, allow one to perform exact and symbolic computation too. We shall explore this new area. It is demonstrated how speed, efficiency and memory usage considerations often lead to surprising and fundamentally different algorithmic solutions in a symbolic or exact context. Applications and examples involve stability of linear systems, model approximation, and linear matrix equations with free parameters. Practical classes serve to demonstrate the techniques and to make the student familiar with exact and symbolic computation.

**Knowledge and understanding:** Students familiarize themselves with state and output feedback to achieve control of dynamical systems. Concretely, they learn to (mathematically) build a basic stabilizing feedback controller for a linear input-output dynamical system, using a combination of different design techniques. Students learn methods for exact numerical and symbolic computation, as used in algebraic computation with unspecified parameters. They also learn in which ways these are different from the more commonly used approximate numerical (floating-point) methods: in terms of accuracy, speed (complexity), and memory usage.

**Applying knowledge and understanding:** Students will be able to construct and implement, for a given linear dynamical input-output system: (a) stabilizing state feedback, (b) full state observer, and (c) additional dynamics to perform tracking of a
specified output trajectory. They will also be able to assess the quality of a controller, regarding an optimal control LQ criterion, and in view of the desired settling time and the trajectory approximation. Students will be able to determine the stability of a given linear dynamical system in an exact and/or symbolic algebraic way. They will also be able to efficiently solve linear systems of (matrix) equations involving symbolic parameters, avoiding pitfalls, which arise from techniques from approximate numerical computation.

**Making judgements:** Students will be able to judge the quality of a feedback design for stabilization (regulation) or tracking. Students will be able to indicate which exact and symbolic computation methods will and will not be useful for a given parameterized problem, regarding speed and memory usage.

**Communication:** Students will be able to motivate the design of a feedback controller, the construction of a trajectory approximation, the design of a full state observer, and the implementation choices of the weights in LQ-design. They will be able to explain the concept of feedback in the area of control. Students can adequately discuss speed and efficiency properties of an algorithm (approximate numerical, exact numerical, symbolic) to specialists and non-specialists.

**Learning skills:** Students will be able to read and interpret basic scientific literature on control theory and on numerical and symbolic computation. They can use Matlab and the Control Toolbox and work out ideas computationally. Students can use some of the exact and symbolic functionality of Mathematica and work out ideas computationally.

**Study material:** Syllabus, provided on eleUM. Handouts.


**Exam:** Written exam by computer in two parts, each having a weight of 50% on the final grade: one midterm take-home exam with Matlab on part 1 (control), one final classroom exam with Mathematica on part 2 (symbolic computation). The resit exam is on both parts of the course in a classroom setting.

**ECTS:** 6

**Algorithms for Data Visualization (Code: KEN4213)**

**Examiner:** Dr. S. Chaplick

**Desired Prior Knowledge:** Data Structures & Algorithms. Discrete Mathematics. Graph Theory

**Prerequisites:** None

**Description:** In our modern world, we are surrounded by data sets in all shapes and sizes. An essential aspect of working with data sets (whether relational, quantitative, etc.) is how they should be presented/visualized. Even for a single data set, different visualizations can be better for different tasks. Moreover, the scale of the data sets often restricts the options available when designing how it should be presented (including the choice of algorithm and the appropriateness of preprocessing/cleaning the data to a “visualizable” scale). This course will provide an overview of the basic theoretical and practical aspects of information visualization with a focus on algorithmic approaches. It includes how to visualize relational data (e.g., graphs/networks) and standard approaches for quantitative data sets such as, projecting high dimensional data to lower dimensions for visualisation (e.g., multi-dimensional scaling and t-distributed stochastic neighbour embedding, etc.). We will also cover some aspects of augmenting visualizations with meta-data such as, labeling nodes/points, weighting relations, and information regarding grouping/clustering.

**Knowledge and understanding:** Students will be able to identify and recount the
common approaches used in information and network visualization. For example, students will be able to identify visualization techniques used in systems/media. Moreover, students have a sufficient understanding of such these approaches to implement them.

**Applying knowledge and understanding:** Students will be able to adapt existing and design new visualization approaches to various problems arising in real-world applications. Students will be able to design efficient implementations of such approaches via standard algorithmic techniques, and to argue about their performance.

**Making judgements:** Students will be able to evaluate under which circumstances different visualization techniques are viable, and to select an appropriate technique to be applied depending on the setting/task.

**Communication:** Students will be able to analytically argue about correctness and performance of the used algorithmic approaches. Students will be able to explain and compare different visualization approaches.

**Learning skills:** By the end of the course, students will be able to autonomously evaluate a new visualization technique as it relates to standard approaches. For example, students will also be able to read and discuss visualization literature beyond the scope of this course (as showcased via a short presentation during the course).

**Study material:** Appropriate study material will be provided throughout the course.

**Recommended literature:**

**Exam:** Written exam (80%) at the end of the course, graded exercises (10%) throughout the course, and a short presentation (10%).

**ECTS:** 6

**Period 1.6**

**Research Project AI 2 & DSDM 2 (Code KEN4131 & Code KEN4231)**

**Coordinator:** Dr. ir. K. Driessens

**Tutors:** Dr. ir. K. Driessens & Dr. G. Schoenmakers

**Desired Prior Knowledge:** None.

**Prerequisites:** None.

**Description:** The research project takes place during the three periods of the semester. Project topics are presented at the start of the semester and assigned to students based on their preferences and availability. The emphasis in the first phase is on initial study of the context set out for the project and the development of a project plan. In the second period, the goal is to start modelling, prototyping and developing. In period 3, the implementation, model and/or experiments set out in the project plan has to be finished and reported on. At the end of period 1 and 2, a progress presentation takes place. The project results in a project presentation, a project report and possibly a public website and/or product.

**Knowledge and understanding:** Students get to know and possibly contribute to state of the art methods within the fields of Artificial Intelligence and/or Data Science for Decision Making to answer an open question.

**Applying knowledge and understanding:** Student write their own research plan in coordination with a staff member (plus possibly outsiders) who act as clients with an
open question. Students with different backgrounds and from both masters work together in teams to build and evaluate an answer to an open question. Students find, judge the suitability, apply, and evaluate state of the art techniques to answer questions and construct applications in the field of Artificial Intelligence and Data Science. Students apply the accumulated knowledge from other educational activities in application specific areas

**Making judgements:** Students judge feasibility of tasks, attainability of goals, and the amount of work involved.

Students think about the possible consequences of their work. Students evaluate state of the art and the applicability and scope of research results.

**Communication:** Students will learn to:
1. orally communicate and cooperate with peers
2. orally report on progress and intermediate results to superiors
3. orally negotiate and communicate with clients
4. communicate their ideas in written form, both for an academic and a general audience
5. give effective presentations

**Learning skills:** Students increase their own level of knowledge in a specialised sub-discipline of the field of Artificial Intelligence and/or Data Science. Students perform research into recent state of the art techniques. Students learn that the field of Artificial Intelligence and Data Science are constantly evolving beyond what is taught in class

**Study material:** Slides provided at the end of joint information sessions. Literature provided by the project supervisors.


**Exam:** Phase 1: project plan + presentation (15%); Phase 2: layman’s website + presentation (15%); Phase 3: Project report + presentation (70%)

**ECTS:** 6

### Period 2.1, 2.2, 2.3

**Electives**

Period 1, 2 and 3 of year two of the master’s program consist of electives to be chosen by the student. This optional program can be assembled at your own choice, but within academic significance and relevance to your master’s track. The Board of Examiners has to evaluate and approve the chosen combination of electives. The electives consist of the following options to choose from: courses to be followed at DKE, at other UM Master programmes, at another university, a research project, an internship, a study abroad at a foreign university, or a project. Note that you have obtained at least 40 ECTS of course year 1 in order to enter the second year of the programme.

**Electives at Maastricht University outside DKE**

It is possible to take electives at other relevant master’s programmes at Maastricht University for at most 13 ECTS in the second year of the programme. The following courses below will be automatically approved by the Board of Examiners of DKE. You should apply through the Special Course Approval procedure via the My UM Portal. **Note** that they may have limited capacity.
School of Business and Economics
Social Choice Theory (ECB4005) 6.5 ECTS
Supply Chain Operations (EBC4016) 6.5 ECTS
Negotiation and Allocations (EBC4193) 6.5 ECTS
Intellectual Property Rights in a Digital Economy (EBC4026) 6.5 ECTS
High-Dimensional Econometric Methods for Big Data (EBC4218) 6.5 ECTS

Faculty of Psychology and Neuroscience
Besides complying that you have passed 40 ECTS, for taking these electives at FPN you should have passed “Advanced Concepts in Machine Learning” and “Autonomous Robotic Systems” at DKE.
Auditory and Higher Order Language Processing (PSY4051) 4 ECTS
Perception and Attention (PSY4052) 4 ECTS
Sensorimotor Processing (PSY4055) 4 ECTS
Exam: Depends on content of the elective program.
ECTS: 30

Period 2.4, 2.5, 2.6

Master’s thesis AI & DSDM (Code: KEN4160 & Code: KEN4260)
The Master’s Artificial Intelligence and Data Science for Decision Making will be completed by writing a master’s thesis. The thesis is produced individually and is the result of a master’s research project that runs during the second semester of year 2 of the master’s programme. In the first phase, the emphasis is on self-study, subject determination, planning and some preliminary research. Then the actual research is started. The final phase is used to finalize the master’s thesis. The master’s project is completed by a presentation of the results. The master’s project will be supervised by one of the senior researchers.
Exam: Master’s thesis and presentation.
ECTS: 30

* Note that when you enrol in February, you follow your electives in period 2.4, 2.5 and 2.6 and work on your master’s thesis in period 2.1, 2.2, 2.3.
2.4.3 **Master’s thesis Artificial Intelligence and Data Science for Decision Making**

At the end of the master’s study, each student has to write a thesis. This thesis has to be written and presented individually, and accompanied by relevant attachments and software. In order to start working on the thesis, a student needs to have obtained at least 60 ECTS (among which are 40 credits of the first year).

**General procedure**

The process of writing a master’s thesis consists of 6 phases. It is scheduled in the last semester of the master’s study. The time frame given below is an indication for these phases.

**Phase 1: Topic selection**

At the end of the previous semester, the students are informed of the main directions of research in the three research groups at the DKE department. Based on this information, students acquire more information about specific possibilities in the areas by means of individual discussions with relevant researchers available. These discussions take place upon the initiative of the student.

**Phase 2: Thesis Research Plan**

At the start of the first period of the semester, each student must have chosen a thesis topic and a principal thesis supervisor. The student creates a thesis research plan, which is to be signed by the student and the thesis supervisor, and then handed over to the master’s thesis coordinator. The plan is sent to the Board of Examiners for approval. In the fourth week of period 1, the students present their research topics during a 10 minutes presentation in front of the DKE staff members and fellow students.

**Phase 3: Research**

During the second period of the semester, the student carries out his/her own research. This research process will be guided by the thesis supervisor through a series of frequent appointments, preferably on a weekly basis. In the fourth week of period 2, the student is invited to present the first research achievements, with audience the DKE staff members and fellow students.

**Phase 4: Writing**

At the end of the second period, the research is finished and the first version of the thesis is submitted to the thesis supervisor. The thesis supervisor, as well as a second assessor, will evaluate the thesis and provide a first reaction within one week. The second and final submission will take place during the second week of the third period.

**Phase 5: Preparation for presentation**

In the third week of the third period, the student prepares a final presentation of the thesis research. This individual presentation will have a maximum length of 30 minutes, followed by 15 minutes of discussion.

**Phase 6: Presentation**

The master’s thesis will be presented in the last week. Two dates are reserved for that occasion. The presentation is open for all students and employees of the bachelor’s Data Science and Artificial Intelligence and master’s programmes, Artificial Intelligence
and Data Science for Decision Making and anyone else who was involved in the thesis research. The final decision on the grade for the master’s thesis will be made shortly after the presentation.

Requirements and assessment
For the master’s thesis research, every student has to conduct a short scientific research. This can be an empirical or a theoretical research. The topic is open, as long as it fits into the field of the master’s program. DKE staff will briefly introduce their main areas of research, but students are encouraged to propose a research topic themselves. The topic and the research question have first to be approved by the thesis supervisor. To achieve this, the student will create a master’s thesis research plan using the form provided by the Board of Examiners that contains amongst others:
• Date;
• Name and student number;
• Name of the thesis supervisor(s) and examiners;
• Title of the master’s thesis, start and end date of the thesis research;
• Short description of the problem statement and research questions;
• A list of the main research activities, with a time schedule.

This plan will be signed by the student and the thesis supervisor and prospective examiners and then handed in to the Board of Examiners for the formal approval. It is possible to execute the master’s thesis research as an external training period. This should be well defined in the master’s research plan. In this case, the plan should also include the name of the company, the name of the external supervisor, the size of the project and any agreements about payment and confidentiality. The plan should also be signed by the external supervisor.

The research needs to be original in such a way that the thesis supervisor is convinced that this research has not been done before. The research also needs enough depth and still it must be possible to finish it in the set amount of time. It is possible for multiple students to cooperate in a research project as long as it is clear who did what. Moreover, every student has to write their own thesis reflecting his part of the research. The assessment will be based on the contents and form of the thesis and the presentation of this thesis.

Content aspects
The thesis describes the problem statement, research questions, approach and results of the research. This has to be done in a clear, structured and scientific manner. This includes:
• a clear introduction in which the problem statement and research questions are presented;
• the master’s student shows proper analysis of complex issues in a new context and is able to formulate a proper problem statement;
• a clear conclusion, based solely on the already used thought out principles and derived results;
• a clear line is shown between problem statements, approach, methods and the derived results;
• a motivation of the followed approach, reflecting on standard methods and their presuppositions,
• an adequate description of the followed approach;
• a purposeful and systematic way of collecting data;
• an honest, clear and concise description of the derived results, if necessary using tables;
• an analysis and discussion of the results;
• the usage of relevant and recent literature for the reasoning in the thesis.
• the correct usage of references.

Design aspects
Correct scientific references have to be used. Images and tables are accompanied by an index and caption. Mathematical formula, definitions, etc. have to be properly designed and numbered. The start and end of mathematical formulae have to be properly defined.

Language aspects
The thesis has to be written in English, considering correct spelling, syntactical structure of sentences and structure of content in paragraphs. The target audience consists of fellow master’s students and lecturers. Any jargon and/or abbreviations have to be explained unless they are common knowledge for this audience.

Citations
It is allowed to use several short citations. These citations have to be clearly referenced and have to be typographically distinguishable (that is, citations are placed in quotes). Non-allowed citations or missing references will result in a non-pass.

2.5 Courses at other faculties or universities
If a student from DKE would like to participate in courses at other faculties or other universities, approval from the Board of Examiners is needed in advance. For more information, please contact the Student Affairs Office or visit the study advisor.

2.6 DKE Honours Programme
DKE offers its talented and top-performing bachelor’s students the possibility to participate in the DKE Honours Programme. This programme offers two variants MaRBle 2.0 and KE@Work.

MaRBle 2.0
In MaRBle 2.0, you will get the opportunity to work on a state-of-the-art research project. Work will be organized in a similar way as in professional research institutes where participants work together as individual experts on a team project. For more information on the MaRBle 2.0 please contact dr. R. Möckel (rico.mockel@maastrichtuniversity.nl)

KnowledgeEngineering@Work (KE@Work)
Students admitted to the KE@Work path are placed at a business through a careful selection and matching process. During the full second and third year of the bachelor program, they spend 50% of the time in class and 50% at the business, where they work on solving academic challenges and complex business problems, under supervision of dedicated business and university supervisors. For more information on KE@Work, please contact Carla Franken, kework@maastrichtuniversity.nl.

Both honours programme takes place in years two and three of your bachelor’s programme. Selection of honours students will happen in the second semester of
year 1. If you successfully complete the honours programme, this will be certified on an honour’s diploma supplement.

2.7 Educational Minor Bachelor DKE

The Educational Minor leads to a limited second-degree teaching qualification. If you have successfully finished the minor, you are qualified to teach at lower secondary schools of VWO, HAVO and VMBO-tl (MAVO) level. Students in the BSc Data Science and Artificial Intelligence can -upon successful completion - acquire a teaching qualification for the main subject of Mathematics. The Educational Minor will be organised in close cooperation with the Fontys Leraren Opleiding (Teacher Training) in Sittard (FLOS) and Tilburg (FLOT). The main language of the Educational Minor is Dutch.

The programme contains several pedagogical-didactic courses in semester 5, along with education aimed at teaching methodology. There is also a mandatory practical internship, in the form of work placements, which is spread out over semester 5 and 6. The education meetings mostly take place at UM and occasionally at the Fontys Leraren Opleiding in Sittard. The practical internship will be done at several secondary schools in the whole province of Limburg and will continue until the end of semester 6. During the practical internship, the student spends one day a week at a secondary school for the course of a full school year. In this way, the necessary teaching experience is obtained. Knowledge and practice are closely connected in the Educational Minor.

Successful completion of the educational minor yields 35 ECTS of which five are extracurricular. This means that these 5 ECTS cannot be used to replace any other components of the original bachelor program.

The minor is UM wide and is also open to students of other UM faculties. Students in the BSc Data Science and Artificial Intelligence who wish to participate in the program should have accumulated, by the end of their second year, all 60 ECTS from the first year components and at least 52 ECTS from the second year components. Prior to their enrolment in this minor, a motivated request for participation has to be submitted in Dutch to the Board of Examiners (dke-exams@maastrichtuniversity.nl). Enrolment is dependent on selection and prior permission of the Board of Examiners.

If you have any questions regarding the contents of this educational minor, then please contact prof. dr. Frank Thuijsman, (f.thuijsman@maastrichtuniversity.nl).
3 Facilities for students

In this chapter, you will get an overview of possible facilities that Maastricht University offers its students.

3.1 Student Affairs Office

The Student Affairs Office, among other things, takes care of the organization and administration of the education.

Visiting address:
Paul-Henri Spaaklaan 1, 6229 GT Maastricht
Postal address: P.O. Box 616, 6200 MD Maastricht, the Netherlands.

Office hours:
PHS, C.1006 daily between 10.00 - 11.00 hrs. and 15.00 - 16.00 hrs.

Contact:
Admissions: boa@maastrichtuniversity.nl, Tel.: +31(0)43 388 26 77
Exam Administration: bo-dke@maastrichtuniversity.nl, Tel.: +31(0)43 388 35 25
Scheduling: scheduling-dke@maastrichtuniversity.nl, Tel.: +31(0)43 388 35 25

3.2 Administrative structure of the Faculty

The administrative structure of the Faculty is laid down in the faculty regulations. The dean is responsible for the faculty’s administration. More information is to be found on the website: www.maastrichtuniversity.nl/dke.
Faculty Board
The Faculty Board, chaired by the dean of the Faculty of Science and Engineering, runs the Faculty. The Faculty Board is charged with the general management and administration, as well as its policy regarding academic research and education.

Faculty Council
The Faculty Council is entitled to submit proposals and present their opinion to the Faculty Board regarding any matters relating to faculty administration, policy, education and research. The Faculty Council has rights of approval, e.g. regarding faculty regulations, research programmes, and the implementation of a binding study advice, and rights of advice, e.g. regarding the budget.

Director of Studies
The directors of studies (prof. dr. Mark Winands for the Master programmes and dr. Pietro Bonizzi for the Bachelor Programme) are responsible for the organization and coordination of all teaching activities. The DKE Education Programme Committee (EPC) advises the directors of studies.

Programme Committee
There is one EPC for the Bachelor’s Data Science and Artificial Intelligence, and the two Master’s programmes in Artificial Intelligence and Data Science for Decision Making. The EPC is responsible for advising the Faculty Board, the Director of Studies and the Board of Examiners. Furthermore, the EPC is entitled to advice in any subject related to the programme, and consists out of eight members, four students and four members of the academic staff. In addition, there are two advisors: the quality assurance officer and the study advisor.
All correspondence for the Programme Committee should be addressed to Dke-secretariat@maastrichtuniversity.nl or by postal mail to:
Department of Data Science and Knowledge Engineering - Maastricht University P.O. Box 616, 6200 MD Maastricht.

Board of Examiners
The Board of Examiners is in charge of the organization and supervision of the examinations and is appointed by the Faculty Board. All correspondence for the Board of Examiners should be addressed to dke-exams@maastrichtuniversity.nl or by postal mail to:
Department of Data Science and Knowledge Engineering - Maastricht University Student Affairs Office, P.O. Box 616, 6200 MD Maastricht.

Board of Admissions for the Master’s Programmes
The Board of Admissions is responsible for granting the admission requests for entering a master’s programme, and is appointed by the Faculty Board. All correspondence for the Board of Admissions should be addressed to boa@maastrichtuniversity.nl or by postal mail to:
Department of Data Science and Knowledge Engineering - Maastricht University Student Affairs Office, P.O. Box 616, 6200 MD Maastricht.
3.3 **Teaching Material**

For each project, a project book is published. The project books and the education schedules of each period are available two weeks before the start of a new period, at the latest. If the prospectus makes no mention of study material (= obligatory literature) or recommended literature, the study material will be mentioned on a separate book list that is available on the course information page in Canvas or at the teacher’s office. All students will receive a letter from the study association Incognito, which can order the books with a discount at the bookshop.

3.4 **Participation in the Education**

The students are expected to be available from Monday through Friday from 08.30 a.m. to 18.00 p.m. for educational activities.

3.5 **Announcements concerning Educational Matters**

Announcements concerning educational matters will be published through the Student Portal. In the hallway of the building there are bulletin boards and TV screens where relevant announcement are also published. General announcements for students will also be published in the newspaper of Maastricht University the Observant. The Observant appears every Thursday. Students are mainly approached through e-mail and through the Student Portal. We advise students to check for new announcements/emails daily.

3.6 **Change of Address Student**

Except for the Student Portal, the bulletin board, the Observant, the Student Affairs Office makes use of mailings to students. Overviews of study results and results of examinations may be sent by regular mail. If there is a change in the study address or the address of the student’s parents, this should immediately be changed in Studielink. Do not forget to mention the commencing date of the change. During the academic year, the student’s study address is considered as their postal address. You may contact boa@maastrichtuniversity.nl for help.

3.7 **Computer Rooms and Project Rooms**

Scheduled practical lectures have priority over private use by students. The room is open to students from Monday through Friday from 08.00 a.m. until 06.00 p.m. Students can use these computers to make exercises, develop software, and communicate with other people outside the institute. The computers are connected to the university network Maasnet. Several software packages have been installed, such as Word, Access, Powerpoint, Matlab, Mathematica and Java. Also specific packages that will be used during the courses have been installed. The computer facilities are managed by the system managers of DKE, which can be contacted by emailing lo-fhs@maastrichtuniversity.nl or by phone +31(0)43-388 54 93.
**Project/meeting rooms**

Scheduled practical lectures have priority over private use of the project rooms by students. The rooms are open to students from Monday through Friday from 08.00 a.m. until 06.00 p.m. Wireless internet is available throughout the whole building. For questions about the system management of the computer room, please refer to the system managers of DKE, tel. +31(0)43-388 54 93 or by mail: lo-fse@maastrichtuniversity.nl.

**House rules for all computer/project/meeting rooms:**
- Users are not allowed to download illegally acquired materials;
- Users are not allowed to illegally download materials;
- Users are not allowed to install illegally acquired software;
- Users should use their own devices for saving data, or save your data on your personal network drive (I:);
- Users should handle the furniture with care;
- It is strictly forbidden to eat or drink in these rooms;
- For the regulation of the air conditioning system, students may contact the Student Support staff.

### 3.8 Faculty Counsellors for Students

**Study Adviser**

The study advisers Tessa Fox and Wendy Brandt are staff members whom you can contact if you have any questions concerning your study and can be reached at telephone number +31(0)43-3885361, in room C.1012 at PHS and through sa-dke@maastrichtuniversity.nl. They are familiar with the organization of the education, the faculty organization and the study. The study adviser is a primary advisor for students. If your study comes to a standstill, for whatever reason, you can contact the study adviser. It is also the right person to talk to if you have any questions to which you cannot find any answers in the faculty prospectus or during faculty information meetings. But also in case of personal circumstances due to which your study or personal life are suffering, for instance illness or family circumstances, your study adviser can help. In a situation like that it is very important to contact the study adviser in time, with respect to a possible appeal by the Regulations on Financial Support for Students. The study adviser may also call up students for a talk if it appears that their results are falling back.

**Internationalization**

For any questions, you may have about studying a semester at a foreign university, or about a practical training abroad, for support, and for direct information you can contact during opening hours DKE’s international relations officer Wendy Brandt via dke-international@maastrichtuniversity.nl.

### 3.9 Student Services Centre (SSC)

The Student Services Centre is responsible for the preparation and execution of the policy of Maastricht University in the area of general student provisions. The Student Services Centre sees to the maintaining of the relationship with new students and alumni, an agreeable living environment for students and student associations, and guidance unrelated to the studies. The SSC publicizes current information in the university magazine Observant and provides extensive information on the Internet.
Students are able to ask questions via the electronic service centre. In short, this department has a number of specialized service units for student-related issues such as accommodation, sports, information on studies and work and career advice. In addition, there is a central information desk in the main entrance hall of the Visitors’ Centre, to which current and prospective students may address their questions. Visiting address: Bonnefantenstraat 2, Tel.: +31(0)43-388 53 88, www.maastrichtuniversity.nl/ssc.

3.9.1 Visitors’ Centre and student registration

**Information Desk**
The information desk in the UM Visitors’ Centre at Bonnefantenstraat 2 is the first point of contact for current and new students. It provides the following services:

- Help with admission and (re)registration;
- Information on and help with visas, scholarships, bank accounts and (health) insurance;
- Changing of address;
- Payment of tuition fees;
- Cancellation of registration;
- Reimbursement of tuition fees;
- Proof of payment/registration;
- Collection of your first UM-card;
- Help with housing;
- Appointments with student deans, psychologists, and career services;
- UM gifts.

Please feel free to contact our Call centre for answers to any of your questions regarding studying at Maastricht University.
Tel.: +31(0)43-388 53 88, e-mail: study@maastrichtuniversity.nl
FAQ: https://ssc.esc.maastrichtuniversity.nl
Opening hours Monday-Friday 08.30 a.m.-18.00 p.m.

**Admissions**
The Admissions Office is responsible for helping prospective bachelor’s and master’s students with their admission to Maastricht University. They provide information on diploma evaluation, admission procedures, sufficiency tests for courses and languages, etc. They are also responsible for the coordination of the prospective students’ admissions audit. The Admission Office works in close cooperation with the Registration, Visa and Scholarship Office to provide the best possible service to prospective students. For any questions on admissions, please contact the Admissions Office by e-mail: study@maastrichtuniversity.nl or contact their Call centre +31(0)43-388 53 88.

**Visa and Scholarship Office**
The Visa and Scholarship Office is responsible for immigration matters and scholarships for prospective and current students. The Visa Office offers prospective and current students assistance with obtaining visas, work or residence permits and with the extension of residence permits. For any questions on visas, please visit our website: www.maastrichtuniversity.nl/visa or e-mail: visa@maastrichtuniversity.nl.
Scholarship Office
Prospective and current students can obtain information about scholarships (Socrates/Erasmus, HSP Huygens, cultural treaties, NFP, UM High Potential and UM Company scholarships). For any questions on visas, please visit our website: www.maastrichtuniversity.nl/scholarships or e-mail: scholarships@maastrichtuniversity.nl.

3.9.2 UM Career Services

UM Career Services aims at assisting students in successful preparation for their future career. This goal is achieved by providing students with the required education, information, advice and counselling. In addition UM Career Services links students of Maastricht University to the job market in various ways.
UM Career Services offers workshops, job Interview simulations, Quick career advise and more intensive counseling. For more information, please see www.maastrichtuniversity.nl/careerservices or contact your study adviser at DKE

3.9.3 Student Guidance

At Maastricht University, you are expected to be independent and take care of all your affairs yourself. This does not mean you will not have any questions! The student counsellors can surely answer many of your questions. Below an overview of the counsellors working at Student Services Centre:

Psychological support (Student psychologists)
Student Psychologists may be consulted in case of personal problems. Examples of complaints and problems include:
• Study related problems like study stress and fear of failure;
• Psychological complaints such as anxiety, depression, eating disorders, stress-related complaints, lack of confidence, dealing with traumatic experiences.
It need not be obvious beforehand what the problem is before an appointment can be made with one of the student psychologists.

The student psychologists can help you by means of individual guidance and/or group training (in Dutch and English). Examples of group training: Training course on fear of failure, Study efficacy Group, Mourning Group, Stress management, Assertiveness training, etc. For more information: www.maastrichtuniversity.nl/studentguidance e-mail: studentenspsychologen@maastrichtuniversity.nl
For making an appointment tel.: +31(0)43-388 53 88 or use the online tool on the website: www.maastrichtuniversity.nl/studentguidance

Study related legal support (Student deans)
Student Deans help you when you have questions about:
Student Deans help you when you have questions about:
• Your rights in case of a study delay because of illness, pregnancy, family circumstances or practising top sports;
• Student grants;
• Studying with a functional impairment;
• Membership of a council, board, committee or membership of the board of a student organization;
• Other questions concerning your rights as a student.
For more information: www.maastrichtuniversity.nl/studentguidance
e-mail: studentenadecanen@maastrichtuniversity.nl
Open visiting hours at the SSC, please check the website for correct timeslots

**Studying with a disability, chronic illness or dyslexia**
It is important to Maastricht University that students with a functional impairment can successfully complete their studies without too much delay. By functional impairment UM means all disorders that are of a permanent or temporary character. Amongst these are all motor, sensory or psychological disorders, but also non-visible disorders, such as dyslexia, chronic illness, physical complaints, depression and the like. The Service desk Disability Management (DM) is available to students (with a functional impairment), prospective students, student counsellors, teachers, parents and others who are interested and offers:
• Information (about studying with impairment, laws, (UM-) regulations and external organizations);
• Advice;
• Support (for example, by arranging facilities);
• Help with requesting (education) facilities;
• Help answering questions about studying with a functional impairment;
• Assistance addressing complaints and problems with regard to this topic.

When you have a functional impairment or if you get confronted with an impairment during your study, this might imply that adjustments and facilities are needed in order to reduce study hindrances and delays. Facilities have to be requested on time. The Service desk closely cooperates with student deans, study advisers, student psychologists and career counsellors.
For more information: www.maastrichtuniversity.nl/disability
e-mail: disability@maastrichtuniversity.nl
Open visiting hours: Monday - Thursday from 11.00 a.m. to 13.00 p.m.
Tel.: +31(0)43-388 52 72.

### 3.9.4 Student Housing

Student housing helps students find accommodation in Maastricht and the surrounding area. All mediation and registration takes place via www.maastrichthousing.com. Most students, who study abroad, opt for a room in the University Guesthouse, and so you only need to contact the Student Services Centre for more information.

### 3.9.5 Studium Generale

Studium Generale is a University department that offers a wide program of lectures and cultural activities on a variety of themes. You can visit the activities from September through May (academic year). The programme of Studium Generale consists of:
• Interesting lectures and exiting debates;
• SG on Stage: comedy, pop music, world music and theatre. You can also perform at the Open Mic, the student singer-songwriter contest and the battle of the bands;
• The SG Science Café, a meeting place for scientists and university students.
An opportunity to debate with scientists in an informal atmosphere. For more information: www.sg.unimaas.nl

3.9.6 Sports

University sports at their best: affordable, relaxing, open, challenging and international. More information visit: www.maastrichtuniversity.nl/lifeum/sports-maastricht

3.9.7 Center for European Studies (CES)

If you are a student who would like to study at Maastricht University for one semester or for a summer program, you can study through the Center for European Studies (CES). You can choose from the wide variety of all courses that the UM offers. CES is there to ensure full integration into university life and help you with all kinds of questions you might have, both academic and practical. If you are looking for a fantastic European experience, individual guidance, 24/7 assistance in case of an emergency and help with anything from an airport pick-up to finding a room, CES is the place to be.

For more Information: www.maastrichtuniversity.nl/ces
e-mail: ces.info@maastrichtuniversity.nl Tel.: +31(0)43-321 52 82.
4 Transnational University Limburg

The transnational University Limburg (tUL) is a cooperation between the universities of Hasselt and Maastricht. Both Master programmes are embedded in the School of Information Technology (SIT) of the tUL. It is possible to follow incidental courses at the campus of the tUL at Hasselt University, Belgium. Students who want to make use of this possibility should individually ask permission to the Board of Examiners of DKE. More information on the content of these courses can be found at: www.uhasselt.be/informatica

5 Education and Examination Regulations 2020-2021 (EER)

Maastricht University, Faculty of Humanities and Sciences, Department of Data Science & Knowledge Engineering

5.1 Bachelor Data Science and Knowledge Engineering

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Section 1 | General provisions

Article 1.1 Applicability of the regulations
These regulations apply to the education and exams and examinations of the bachelor’s programme Data Science and Knowledge Engineering (hereinafter to be referred to as: ‘the programme’) and to all students who are registered for the programme.
The programme is provided by the Faculty of Science and Engineering, hereafter called the faculty, at the Department of Data Science & Knowledge Engineering, hereafter to be called the department.
The regulations were adopted by the faculty board after advice and consent from the programme committee and after consent from or in consultation with the faculty council.
The regulations will take effect on 1 September 2020 for the 2020-2021 academic year.
These regulations also apply to students from other programmes, faculties or institutions of higher education, insofar as they follow components of the programme to which these Education and Examination Regulations apply.
For components of the programme that students follow at another degree programme, faculty or institution of higher education, the Education and Examination Regulations for the other programme, faculty or institution apply to the component in question.

Article 1.2 Definitions
In these regulations, the following definitions apply:
a. The Act: the Higher Education and Scientific Research Act (Wet op het hoger onderwijs en wetenschappelijk onderzoek);
b. Student: a person who is registered at the university for education and/or to take exams and the examination of the programme;
c. Course: a study unit of the programme within the meaning of the Act;
d. Propaedeutic phase: the initial period for the programme with a study load of 60 credits, coinciding with course year 1;
e. Course year: year 1, year 2 or year 3 of the programme;
f. Academic year: the period from 1 September of a calendar year up to and including 31 August of the following calendar year;
g. Programme: the bachelor’s programme referred to in Article 1.1 of these regulations, consisting of a coherent whole of study units;
h. Exam: a component of the examination as referred to in Article 7.10 of the Act;
i. Practical / Skill: practical exercise as referred to in Article 7.13(2)(d) of the Act, in one of the following forms:
- writing a thesis;
- carrying out a (group) project;
- performing a research assignment;
- developing a software program;
- writing a paper, creating a technological design or performing another written assignment;
- participating in field work or a field trip;
- completing an internship;
- participating in an activity intended to develop certain skills;

j. Project Skill: practical training associated as part of the project.

k. Examination: the final examination for the bachelor’s programme;

l. Credit: a unit expressed in ECTS credits, with one credit equalling 28 hours of study;

m. Board of Examiners: the board referred to in Article 7.12 of the Act;

n. Programme Committee: the representation and advisory body that carries out the duties described in Article 9.18 and 9.38c of the Act;

o. Examiner: the person designated by the Board of Examiners to administer exams and to determine the results of such exams;

p. Faculty Board: the faculty board of the Faculty of Science and Engineering, as referred to in Article 9.12 of the Act;

q. Negative Binding Study Advice: the advice in accordance with Article 7.8b of the Act entailing that the student cannot continue in the programme;

r. Semester: part of an academic year, either starting first of September and running for 20 educational weeks, or starting first of February running for 21 educational weeks;

s. Block: part of a semester during which educational activities take place;

t. KE: Data Science and Knowledge Engineering;

u. DKE: Department of Data Science & Knowledge Engineering;

v. UM: Maastricht University;

w. BSA Committee: the committee that issues the (negative) Binding Study Advice on behalf of the Faculty Board;

x. Student Handbook: the programme guide, which includes further details about programme specific provisions and information.

The other terms have the meaning given to them by the Act.

Section 2 | Admissions

Article 2.1 Matching
Participation in matching is an optional part of the admission procedure. The matching process starts with filling out an online questionnaire. A member of the academic staff and the study adviser will assess the answers and give their feedback via email. Based on the outcome, the prospective student may receive an invitation to come in for a personal meeting with a member of the academic staff and the study adviser.

Article 2.2 Pre-university education requirements
A person will be granted admission to the programme if he/she has a pre-university education diploma referred to in Article 7.24 of the Act with the pre-university education profile having WI B; or if he/she has been exempted from this under the Act. Equivalent non-Dutch diplomas are referred in Appendix 2.
Persons who do not have a diploma with the pre-university education profile having WI B, which grants admission to the programme, but who have an equivalent diploma which grants admission to the programme under the Act, may register for the programme only after demonstrating that he or she has sufficient knowledge on the level of the final pre-university education examination of the following subject of the required pre-university education profile: Mathematics B.

**Article 2.3 Language requirement with non-Dutch diplomas**

a. Holders of a non-Dutch diploma can only register if they have met the minimum English language requirement corresponding to IELTS (international English Language Testing System) with a score of at least 6.0

b. The requirement referred to under (a) is met if the person concerned has obtained one of the following diplomas or certificates:
   - A secondary education diploma issued in an EU/EEA country where the person concerned has followed English up to and including the final year;
   - A diploma issued in an non-EU/EEA country that is at least equivalent to a VWO diploma and where English is the official language of instruction in education;
   - A completed bachelor’s or master’s study programme where the language of instruction is English;
   - an International or European Baccalaureate, a US high school diploma or UK GCE A-levels, or
   - Can demonstrate sufficient proficiency in English, for example through English taught courses, internships or work experience in an English environment, or
   - can submit one of the following language test certificates.
     - IELTS (6.0 minimum)
     - TOEFL Paper-based test (550 minimum)
     - OETL Internet test (80 minimum)
     - TOEFL Computer test (213 minimum)
     - TOEIC (670 minimum) the modules ‘listening and reading’ and ‘speaking and writing’ must be completed successfully
     - Cambridge [First Certificate in English (FCE) Grade B, First Certificate in English (FCE) Grade C] or
     - similar accredited certification

**Article 2.4 Entrance examination (Colloquium Doctum)**

1. A person who does not meet the prerequisites referred to in Articles 2.1 and 2.2 can take part in an entrance examination (colloquium doctum), in accordance with Article 7.29 of the Act.
2. A person who wishes to sit the colloquium doctum must be aged twenty-one or over on the date the prospective program starts. This requirement may be waived if the person in question holds a diploma issued outside the Netherlands that grants admission in the country of origin to a programme at a higher education institution. The age requirement can also be waived if the person in question has refugee status and cannot present his/her diploma for this reason.
3. The entrance examination referred to in Article 7.29 of the Act concerns the following subjects at the level indicated:

   - Mathematics: Performing a number of exercises based on the subjects that are discussed in some chapters of a book to be specified later. These exercises are given in English.
4. The Board of Examiners may lay down further instructions in the Rules and Regulations.
5. The Board of Examiners is responsible for conducting this examination.

Section 3 | Contents and Structure of the Programme

Article 3.1a Objectives of the programme
The goals of the programme are as follows:
1. To educate students at an academic level in Data Science and Knowledge Engineering based upon Mathematics, Computer Science, and Artificial Intelligence.
2. To teach students how to analyse and to solve both Data Science and Knowledge Engineering problems in a variety of application domains.
3. To prepare students to work in teams by organizing the education in Data Science and Knowledge Engineering according to the Project-Centred Learning system.
4. To prepare students for continuing their studies, in particular the Master programmes in Artificial Intelligence, Data Science for Decision Making and Computer Science offered by the transnational University Limburg (tUL), or for a career in the professional field (IT industry or IT-related application domains).
5. To stimulate students to acquire an international academic orientation.

Article 3.1b Qualifications of the programme
The 29 qualifications of the programme are as follows:

I. Knowledge and understanding
The recipient of a Bachelor of Science degree in Data Science and Knowledge Engineering should have:
1. Basic understanding of key areas in Data Science;
2. Advanced knowledge of a specific area in Data Science up to a level that without further requirements grants access to a Master programme in this area;
3. Basic understanding of key areas in Artificial Intelligence;
4. Advanced knowledge of a specific area in Artificial Intelligence up to a level that without further requirements grants access to a Master programme in this area;
5. Basic understanding of key areas in Computer Science and in Applied Mathematics;
6. Advanced knowledge of a specific area in Computer Science and in Applied Mathematics up to a level that without further requirements grants access to a Master programme in this area.

II. Applying knowledge and understanding for problem solving and design
The recipient of a Bachelor of Science degree in Data Science and Knowledge Engineering should have at least the following thirteen abilities:
7. The ability to understand, apply, formulate, and validate models from the domains of Data Science and Artificial Intelligence;
8. The ability to extract information from data, to interpret results and to convey them;
9. The ability to apply knowledge from the key areas of Artificial Intelligence;
10. The ability to apply the support modules for Artificial Intelligence;
11. The ability to apply methods and tools from applied mathematics and operations research in particular;
12. The ability for constructing and evaluating mathematical and computational methods for a range of application domains;
13. The ability to submit an argument in the exact sciences to critical appraisal;
14. The ability to think analytically and critically, and to apply logical reasoning;
15. The ability to cooperate in a group and to participate effectively as an academic professional;
16. The ability to create an effective project plan for solving a Data Science and/or Artificial Intelligence problem in a supervised context;
17. The ability to apply Data Science and Artificial Intelligence methods and techniques in a business-related practice;
18. The ability to transpose academic knowledge and expertise in a variety of application domains;
19. Readiness to address new problems in new areas, emerging from scientific and professional fields.

III. Making judgments
The recipient of a Bachelor of Science degree in Data Science and Knowledge Engineering should have:
20. The ability to review critically (a) results, (b) arguments, and (c) problem statements from accepted perspectives in the field of Data Science and Artificial Intelligence;
21. A reasonable level of competence in searching and critically processing the professional literature in Data Science and Artificial Intelligence;
22. A reasonable familiarity with the standards of academic criticism;
23. An awareness of, and responsibility for ethical, normative and social consequences of developments in science and technology, particularly resulting from Data Science and Artificial Intelligence.

IV. Communication
The recipient of a Bachelor of Science degree in Data Science and Knowledge Engineering should have:
24. Academically appropriate communicative skills, i.e., the ability to (a) communicate ideas effectively in written form and through the use of Information and Communication Technology, (b) give effective oral presentations, both formally and informally, and (c) understand and offer constructive criticism of the presentations of others;
25. International communication skills;
26. Elementary effectiveness in leading group-wise communication.

V. Learning skills
The recipient of a Bachelor of Science degree in Data Science and Knowledge Engineering should be able to:
27. Reflect on (a) one’s own style of thought, (b) one’s own working methods, and (c) one’s own readiness to take the necessary corrective action;
28. Recognize the need for continued learning throughout a professional career;
29. The ability to manage one’s own learning and development.

Article 3.2 Form of the programme
This is a full-time programme. The programme commences once a year in September.

Article 3.3 Language of instruction
1. The programme is given in English (accountability for this language of instruction is given in Appendix 4). Components of the programme may be in Dutch or in another common language in the EU.
2. Dutch, English, French and/or German texts may be used in the education and exams in the programme.
Article 3.4 Communications and announcement of decisions
1. The Faculty Board, the Board of Examiners, the DKE Management Team and the examiners may use the Student Portal and email via the UM account for communications relating to the education and examinations.
2. The Faculty Board, the Board of Examiners, the DKE Management Team and the examiners may use My UM and e-mail via the UM account to announce decisions.
3. The student must regularly check his/her university e-mail address, the DKE website and the digital learning environment. Information disseminated via e-mail, the digital learning environment or the website will be assumed to be known.

Article 3.5 Study load
The programme has a study load of 180 credits (ECTS), with one credit equalling 28 hours of study.

Article 3.6 Content of the programme
1. The programme includes the following components with the stated study load:

<table>
<thead>
<tr>
<th>Year 1</th>
<th>Course</th>
<th>ECTS</th>
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</thead>
<tbody>
<tr>
<td></td>
<td><strong>Block 1.1</strong></td>
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<tr>
<td></td>
<td>Introduction to Data Science &amp; Artificial Intelligence</td>
<td>4</td>
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<tr>
<td></td>
<td>Introduction to Computer Science 1</td>
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<td></td>
<td>Discrete Mathematics</td>
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<td></td>
<td>Project 1-1</td>
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<td></td>
<td><strong>Block 1.2</strong></td>
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<tr>
<td></td>
<td>Computational and Cognitive Neuroscience</td>
<td>4</td>
</tr>
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<td></td>
<td>Introduction to Computer Science 2</td>
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<tr>
<td></td>
<td>Linear Algebra</td>
<td>4</td>
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<tr>
<td></td>
<td>Project 1-1</td>
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<td></td>
<td><strong>Block 1.3</strong></td>
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<td></td>
<td>Project 1-1</td>
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<td></td>
<td><strong>Block 1.4</strong></td>
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<tr>
<td></td>
<td>Calculus</td>
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<tr>
<td></td>
<td>Data Structures and Algorithms</td>
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<td></td>
<td>ICT and Knowledge Management</td>
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<td></td>
<td>Project 1-2</td>
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<td></td>
<td><strong>Block 1.5</strong></td>
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<tr>
<td></td>
<td>Numerical Mathematics</td>
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<td></td>
<td>Software Engineering</td>
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<td></td>
<td>Logic</td>
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<td></td>
<td>Project 1-2</td>
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<td></td>
<td><strong>Block 1.6</strong></td>
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<td></td>
<td>Project 1-2</td>
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<td><strong>Block 2.1</strong></td>
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<tr>
<td></td>
<td>Databases</td>
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<tr>
<td></td>
<td>Graph Theory</td>
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<td></td>
<td>Probability and Statistics</td>
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<td></td>
<td>Project 2-1</td>
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<td></td>
<td><strong>Block 2.2</strong></td>
<td></td>
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<tr>
<td></td>
<td>Reasoning Techniques</td>
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<tr>
<td></td>
<td>Machine Learning</td>
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<tr>
<td></td>
<td>Linear Programming</td>
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<td></td>
<td>Project 2-1</td>
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<td></td>
<td><strong>Block 2.3</strong></td>
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<td></td>
<td>Project 2-1</td>
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<td></td>
<td><strong>Block 2.4</strong></td>
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<tr>
<td></td>
<td>Human Computer Interaction and Affective Computing</td>
<td>4</td>
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<td></td>
<td>Theoretical Computer Science</td>
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### Year 3 Courses

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<tr>
<th>Block 3.1</th>
<th>Course</th>
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<tbody>
<tr>
<td></td>
<td>Digital Society*</td>
<td>4</td>
</tr>
<tr>
<td></td>
<td>Game Theory*</td>
<td>4</td>
</tr>
<tr>
<td></td>
<td>Semantic Web*</td>
<td>4</td>
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<tr>
<td></td>
<td>Robotics and Embedded Systems*</td>
<td>4</td>
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<tr>
<td></td>
<td>Prolog*</td>
<td>4</td>
</tr>
<tr>
<td></td>
<td>Computer Security*</td>
<td>4</td>
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<td></td>
<td>Project 3-1</td>
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<table>
<thead>
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<th>Block 3.2</th>
<th>Course</th>
<th>ECTS</th>
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<tbody>
<tr>
<td></td>
<td>Large Scale IT and Cloud Computing*</td>
<td>4</td>
</tr>
<tr>
<td></td>
<td>Logic for Artificial Intelligence*</td>
<td>4</td>
</tr>
<tr>
<td></td>
<td>Software &amp; Systems Verification*</td>
<td>4</td>
</tr>
<tr>
<td></td>
<td>Parallel Programming*</td>
<td>4</td>
</tr>
<tr>
<td></td>
<td>Introduction to Bio-Informatics*</td>
<td>4</td>
</tr>
<tr>
<td></td>
<td>Quantum Computing*</td>
<td>4</td>
</tr>
<tr>
<td></td>
<td>Project 3-1</td>
<td></td>
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<th>Course</th>
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<tr>
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<tr>
<td></td>
<td>Operations Research Case Studies</td>
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<td>Intelligent Systems</td>
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<th>Course</th>
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</thead>
<tbody>
<tr>
<td></td>
<td>Bachelor Thesis</td>
<td>18</td>
</tr>
</tbody>
</table>

*Electives (Third year students choose 3 elective courses in each block (3.1 and 3.2)).

In case students have passed both electives of block 2.5, either the course Introduction to Image & Video Processing or Natural Language Processing can replace 1 of the third year electives

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2. After obtaining permission from the Board of Examiners, in blocks 3.1, 3.2, and 3.3, students can also choose to (1) study abroad for a semester, (2) to study elective courses at other UM bachelor programmes of at most 18 ECTS (3) the minor Entrepreneurship or (4) the educational minor. These options cannot be combined and cannot be taken extracurricular unless stated otherwise.

**Article 3.7 Projects and Bachelor’s Thesis**
The programme includes projects and thesis work for which additional regulations apply as specified in the Rules and Regulations.

**Article 3.8 Minor Entrepreneurship**
1. The student has the possibility to choose the minor Entrepreneurship from SBE in blocks 3.1, 3.2 and 3.3 of the third year to fulfil part of the elective obligation in year 3.
2. The minor Entrepreneurship has a total study load of 26 ECTS.
3. Every minor proposal has to be approved by the Board of Examiners.

**Article 3.9 Educational Minor**
1. Additionally, to the possibilities stated in Article 3.8, the student has the possibility to choose an educational minor in year 3 that replaces the elective period in blocks 3.1, 3.2, and 3.3 of the third year.
2. This educational minor has a study load of 35 ECTS of which 5 are extracurricular, hence do not count towards the study load referred to in Article 3.5.
3. For this minor it is required that the student is fluent in Dutch, is starting the minor no later than his/her third year of enrolment of the programme and before starting the minor has, in addition to study progress as mentioned in Article 5.3.2, obtained 52 ECTS from year 2.
4. The Board of Examiners can impose further conditions and restrictions.
5. Participation is only possible with prior permission of the Board of Examiners based on a motivation letter in Dutch, study progress and suitability.

**Article 3.10 Study Abroad**
1. Students can apply to study abroad for a semester, at another University with whom Maastricht University has an Agreement of Exchange.
2. Nomination is decided on by the Board of Examiners based on study progress as mentioned in Article 5.3.2 and motivation of the student;
3. This study abroad will take place in block 3.1, 3.2 and 3.3 and has a study load of 30 ECTS.
4. The selected course programme has to be approved by the Board of Examiners.
5. Further regulations can be found in Appendix 3.

**Article 3.11 KE@Work**
1. KE@Work is an honours programme during the second and third year.
2. Students selected by the KE@Work coordinator on the basis of academic performance and in accordance with the Rules and Regulations, can participate in KE@Work. Students perform Data Science & Knowledge Engineering related tasks as part of an internship for companies selected by the KE@Work coordinator.
3. KE@Work participation has to be approved by the Board of Examiners.
4. An examiner that will act as supervisor will be appointed for every KE@Work internship.
5. KE@Work replaces each of the 3 projects of 6 ECTS in year 2 and 3 by individual projects of 6 ECTS at the selected company. Additionally, the students dedicate their bachelor thesis to a topic relevant for the internship company.
6. Every semester the study performance of the KE@Work student is evaluated. In case study progress suffers, the student is transferred to the educational programme stated in Article 3.6.
7. Additional regulations apply as specified in the Rules and Regulations.

**Article 3.12 MaRBLe 2.0**
1. The programme includes a research-based component called MaRBLe 2.0 (Maastricht Research Based Learning (MaRBLe), which is an honours programme.
2. Students selected by the MaRBLe coordinator on the basis of academic performance and in accordance with the Rules and Regulations, can participate in a state-of-the-art research project.
3. Participation has to be approved by the Board of Examiners.
4. For each project two or more examiners will be appointed.
5. MaRBlE 2.0 replaces each of the 2 projects of 6 ECTS in year 2 by individual projects of 6 ECTS. MaRBlE 2.0 concludes with a research-oriented bachelor thesis.
6. Students can optionally replace the project of 6 ECTS in year 3 by a (separate) individual research project of 6 ECTS.
7. Every semester the study performance of the participating student is evaluated. In case study progress suffers, the student is transferred to the educational programme stated in Article 3.6.
8. Additional regulations apply as specified in the Rules and Regulations.

Article 3.13 Flexible programme and flexible bachelor’s
1. A student registered for one of the faculty’s programmes may, under certain conditions, formulate an educational programme of his/her own which is different from the educational programme stated in Article 3.6. The composition of such a programme must be approved beforehand by the Board of Examiners.
2. The flexible programme must have a study load of 180 credits.
3. In granting the permission, the Board of Examiners will indicate which programme offered by the faculty will include the programme formulated by the student for purposes of the Education and Examination Regulations.

Article 3.14 The examination
The examination consists of the following components
a. Course year 1, offered from September 2013 or later:
   exams from blocks 1.1, 1.2, 1.4, 1.5
   projects 1-1 and 1-2
b. Course year 2, offered from September 2013 or later:
   exams from blocks 2.1, 2.2, 2.4, 2.5
   projects 2-1 and 2-2, or completion of semesters 2-1 and 2-2 of a KE@Work / MaRBlE project
c. Course year 3, offered from September 2013 or later:
   Semester 1: exams from 6 optional courses during blocks 3.1 and 3.2 and the project 3-1. Alternative options in the elective semester are (as further specified in section 3):
   1. A study abroad semester with one of our partner universities;
   2. Electives from other UM bachelor programmes with a maximum of 18 ECTS. They are complemented by elective courses from this programme.
   3. A minor as specified in paragraph 3.8, completed with an elective course at DKE or the minor as specified in paragraph 3.9;
   4. Participation in KE@Work / MaRBlE project (Latter option if extended with permission of the Board of Examiners).
   Semester 2: exams from block 3.4 and the Bachelor thesis.

Section 4 | Education

Article 4.1 Courses; composition; actual design
1. For the programme components, courses are given with the study load stated in Article 3.6.
2. The education is given in the form of classes, project groups, practical training, lectures, individual supervision, or otherwise. On average, the student has 18 hours of face-to-face time per week, but this can differ per period. For further details, please see the student handbook.
3. The educational programme includes 41 weeks per year. During this period students are expected to be available for educational activities.

Article 4.2 Entrance requirements; Prior knowledge
1. The student may only participate in the following components after he/she has passed the listed components (indicated as prerequisites in the student handbook):
   - Year 1 Project 1: after attainment of two out of four courses from the set: Discrete Mathematics, Linear Algebra, Computer Science I and Computer Science II.
   - Year 2 Project 1: after attainment of Project 1-1 and 2 out of 3 of Computer Science I, Computer Science II, and Data Structures & Algorithms.
   - Year 3 Project 3: after attainment of Project 2-1.
2. Subject to the provisions in the first paragraph, the desired prior knowledge to successfully participate in each course is indicated in the student handbook.

Article 4.3 Course registration
The student may participate in a course after he/she has timely registered for it through the Student Portal.

Article 4.4 Attendance and best-efforts obligation
1. The student must participate in at least 70% of the scheduled meeting (e.g. lectures and practical training) of each course in year 1.
2. A student who fails the attendance requirement for a component in year 1 is excluded from the resit exam of that component.
3. Attendance and participation at project skill trainings and project meetings is mandatory. In addition, each student is expected to participate actively in doing tasks with respect to the project and to cooperate actively with their group in order to successfully finish the project assignment.
4. Students who have not met the requirements in paragraph 3 and/or students whose absence or inactivity during the project has been marked as inexcusable by the project coordinator and/or students that have a substandard contribution to the group work will not receive a pass for the project concerned.
5. Attendance and participation in other education activities may be part of an exam when announced in the student handbook or student portal. Prior approval of the Board of Examiners is required.
6. For resits that are offered for practical assignments, the examiner can require that the student has made a reasonable attempt at the regular opportunity for that assignment. This requirement must be communicated upfront to the students on student portal. The Board of Examiners can impose further restrictions in the Rules and Regulations.

Article 4.5 Participation
The blocks mentioned in Appendix 1 are available to a limited number of students enrolled at UM, provided that the students enrolled in the programme will anyhow be placed in the blocks belonging to the compulsory part of their programme, in compliance with the periods the faculty designated to them. For blocks belonging to the optional parts of the bachelor’s programme, most places will be reserved for students enrolled in the programme.
Article 4.6 Practical training
Some courses indicated in Article 3.6 include a practical training for which additional regulations apply as specified in the Rules and Regulations.

Section 5 | Assessment

Article 5.1 General
1. During a course, the student will be tested for academic training and the extent to which the student has sufficiently achieved the stated learning objectives.
2. The student handbook describes the achievements the students must make to pass the course and the criteria on which the student is assessed. Any amendments are published on student portal.
3. The Rules and Regulations describe the assessment procedure.

Article 5.2 Marks
1. Marks are awarded on a scale of 1 to 10. Whole grades are awarded to exams. Both half and whole grades can be awarded for study components of at least 10 ECTS as well as for projects.
2. The student must receive a final mark of 6.0 or higher to pass the component.
3. The highest result attained determines the final mark.
4. NG (no grade) can be assigned as a result of plagiarism or academic dishonesty; or when assessment is incomplete and no grade can be assigned. An NG automatically constitutes a fail and no credit is awarded.

Article 5.3 Order of exams
1. If the student has obtained at least 45 credits in course year 1, he/she may take the exams in course year 2.
2. If the student has obtained at least 60 credits in course year 1 and at least 40 credits in course year 2, he/she may take the exams in course year 3.
3. If the student has obtained at least 60 credits in course year 1, at least 40 credits in course year 2, and at least 140 ECTS overall, he/she may start the bachelor thesis in course year 3.
4. The student may not take an exam for certain components until the entrance requirements as stated in Article 4.2 have been fulfilled.
5. In conformance with article 7.30 paragraph 3 of the Act, the Board of Examiners may grant a student permission to sit other exams than referred to in paragraphs 1, 2 and 3.
6. If a student deviates from the sequence as described in paragraphs 1 through 4, without permission from the Board of Examiners, the result of the component in question can be declared invalid.

Article 5.4 Scheduling and frequency of the exams
1. Students can take exams twice per academic year on dates to be determined by the Board of Examiners: once during or directly after the block (first sit for the exam) and once before the end of the semester in question (resit option).
2. Once a student successfully passes an exam, he/she can re-sit the exam upon permission from the Board of Examiners.
3. In exceptional cases, the Board of Examiners can decide that an exam may be taken at another time than determined in accordance with the first paragraph.
Article 5.5 Registration for exams
1. The student may take an exam for a course for which he/she is registered, after he/she has timely registered for the exam through the Student Portal.
2. For year 1 the resit-exam concerned is only available for students who have complied with the compulsory attendance requirement as stated in Article 4.4.
3. In exceptional cases, the Board of Examiners may, at the student’s request, exempt the student from the obligation in Article 4.4.
4. If a student deviates from the conditions as described in paragraphs 1 through 3, the result of the component in question will be declared invalid.

Article 5.6 Form of the exams
1. In principle, the exams are written. ‘Written exams’ also include taking these exams by computer.
2. In principle all exams and assignments with the exception of projects are on an individual basis, unless explicitly announced otherwise.
3. Oral exams can only take place upon prior approval by the Board of Examiners.
4. The form of the exam is announced by the examiner at the start of the course.
5. Upon request, students with a disability may take exams in a manner which accommodates their specific disability as much as possible. If necessary, the Board of Examiners will obtain expert advice where necessary from DKE’s study adviser and/or the student dean at the Student Services Centre (SSC) before taking a decision in such matters.

Article 5.7 Oral exams
1. Oral exams are taken only by one person at a time, unless the Board of Examiners decides otherwise.
2. An oral exam is given by the examiner in the presence of a second examiner, unless the Board of Examiners has decided otherwise.
3. Oral exams take place in public, unless the Board of Examiners or the relevant examiner decides otherwise in special cases.

Article 5.8 Assessments in exceptional cases
A student can submit a request to the Board of Examiners for an individual assessment
1. This request may be granted if the student has not passed the exam in question due to exceptional circumstances and not granting an individual assessment would result in an unacceptable study delay.
2. The following criteria apply to the granting of an individual assessment for the final component of the programme:
   - It must be the final study result to be obtained.
   - The study delay in case the individual assessment is not granted must be at least one semester.
   - The student must have taken part in the last two regular exam opportunities for the exam for which he/she is requesting another assessment.

Article 5.9 Practicals and written assignments
1. The Board of Examiners may draw up guidelines for the practicals, which include group projects, internships and theses. The guidelines will be included in the Rules and Regulations.
2. The bachelor’s thesis project will be evaluated by at least two examiners (the relevant supervisor and a second evaluator), at least two of them are affiliated with the programme for which the student is registered.
**Article 5.10 Determination and announcement of exam results**

1. The Board of Examiners determines the standards for assessing each examination component. The standards are included in the Rules and Regulations.
2. The examiner determines the result of a written exam within 15 working days of the date on which it was taken and at least 10 working days before the resit, and provides the Education Office with the necessary information to apprise the student of the result.
3. The examiner determines the result of an oral exam within 24 hours and issues the relevant certificate to the student. If more than one student takes the same exam after each other, this period may be extended by up to five working days.
4. When the result of a written exam is announced, it will be indicated how the student can inspect the exam and file an appeal as referred to in Article 6.6.

**Article 5.11 Right of inspection**

1. Within 10 working days of the date on which the result of a written exam, including a computer-based exam, is announced, students may inspect their evaluated work.
2. Within the period referred to in paragraph 1, any interested party may, upon request, inspect the questions and assignments for the written exam and, if possible, the standards based on which it was assessed.

**Article 5.12 Period of validity**

1. Exams which have been passed are valid for an unlimited period. Contrary to the above the Board of Examiners may require the participant to take an additional or replacement exam or examination component for an exam that was passed more than six years ago if the student's knowledge or insight that was examined is demonstrably outdated or the skills that were examined are demonstrably outdated.
2. If exceptional circumstances apply as referred to in Article 7.51 paragraph two of the Act, the period of six years in paragraph one will be extended by the duration of the financial support the student receives from the profiling fund.
3. Sub-tests and assignments that were passed within a component, which was not passed, will lose their validity after the academic year in which they were passed unless the Board of Examiners states otherwise.

**Article 5.13 Retention period for exams**

1. The exercises, answers and the evaluated work of the written exams will be retained in paper or digital form for two years after the exam/examination result is determined.
2. The bachelor thesis and its evaluation will be kept for at least seven years after the evaluation.

**Article 5.14 Exemption**

The Board of Examiners may, at a student's request and having heard the relevant examiner, grant the student an exemption from taking an exam if he/she demonstrates that he/she previously:

1. Either passed an exam for a university or higher professional education programme which was similar in terms of content and level or gained sufficient knowledge and skills relevant to the exam concerned, either through work or professional experience.
2. An exemption may only pertain to an entire course and not a part thereof.
3. At most 60 credits for the programme may be earned based on the exemptions granted.
4. The bachelor’s thesis is excluded from this exemption option.
5. The Board of Examiners will not grant any exemption based on exams passed by a student outside the programme during the period in which the student was barred by the Board of Examiners from taking exams for the programme because of fraud.
6. The same period of validity applies to exemptions as to exam results.

Article 5.15 Fraud
1. ‘Fraud’, including ‘plagiarism’, means actions or omissions by a student which make it impossible in whole or in part to properly evaluate his/her knowledge, understanding and skills. Allowing and/or enabling other students to engage in fraud is also considered fraud under these regulations.
2. ‘Plagiarism’ means the presentation of ideas or words from one’s own or someone else’s sources without proper acknowledgment of the sources.
3. If the Board of Examiners determines that a student has engaged in fraud with respect to an exam or a part thereof, the Board of Examiners can take appropriate measures.
4. In serious cases of fraud, the Board of Examiners can propose to UM’s Executive Board that the student(s) concerned be permanently deregistered from the programme.
5. The Rules and Regulations include further provisions about what constitutes fraud and which disciplinary measures the Board of Examiners can impose.

Article 5.15a Invalid exam
If an exam involves irregularities that make it impossible to accurately assess a candidate’s knowledge, insight and skills, the Board of Examiners may declare the exam invalid for the examinee and/or a group of examinees.

Article 5.16 Unsuitability (Iudicium Abeundi)
1. In exceptional cases and after careful consideration of the interests involved, the Board of Examiners or the Dean/the Faculty Board may ask the executive board to terminate or, as the case may be, refuse the enrolment of a student in a programme, if that student, through her/his behaviour or opinions ventured, has demonstrated his/her unsuitability for the practice of one or more professions for which s/he is trained by the programme s/he follows, or, as the case may be, for the practical preparation for the practice of the profession.
2. The relevant clauses of Maastricht University’s Enrolment Provisions apply.

Section 6 | Examination

Article 6.1 Examination
1. The Board of Examiners determines the result and date of the examination and issues the certificate as referred to in Article 6.3 as soon as the student has satisfied the requirements for the examination programme.
2. Prior to determining the result of the examination, the Board of Examiners may conduct their own investigation of the student’s knowledge regarding one or more components or aspects of the programme if and insofar as the results of the relevant exams/assessments give reason to do this.
3. To pass the examination, the student must pass all components referred to in Article 3.14.
4. To pass the examination and receive the certificate, the student must also have been registered for the programme during the period that the exams were taken.
5. A certificate may only be issued after it has been shown that the student has satisfied all the obligations, including paying the tuition fees.

6. The last day of the month in which the student satisfied all the examination obligations will be considered the examination date (graduation date).

7. Students who have passed the examination and who are entitled to the issuance of a certificate may, stating reasons, ask the Board of Examiners not to do this yet. This request must be submitted at least one month before the final assignment is turned in or the final exam is taken.

   The Board of Examiners in any event grants the request:
   - if the student is selected by the faculty for a double degree, an extracurricular internship or an extracurricular exchange, or
   - if the student holds or will hold a board position for which at least nine months of financial support is awarded from the profiling fund or holds or will hold an ‘INKOM’ board position.

   The Board of Examiners may grant the request if refusal would result in an exceptional case of extreme unfairness because of the fact that the student concerned could not have taken the automatic graduation into account when he/she was planning his/her study.

Article 6.2 Degree

Students who have passed the examination will be awarded the degree ‘Bachelor of Science in Data Science & Knowledge Engineering’.

Article 6.3 Certificate and statements

1. As proof that the examination was passed, the Board of Examiners issues a certificate, after it has been stated by or on behalf of UM’s Executive Board that the procedural requirements for receiving the certificate have been met. The certificate is based on the model that UM’s Executive Board has adopted. One certificate will be issued per programme, even if the student completes several programmes.

2. The certificate that the examination has been passed also indicates:
   a. the name of the institution;
   b. the name of the programme;
   c. the examination components;
   d. (if applicable) the right to practice a specific profession which is related to the certificate;
   e. the degree awarded;
   f. the date on which the programme was most recently accredited or was subjected to the new programme test.

3. Students who are entitled to the issuance of a certificate may, stating reasons, ask the Board of Examiners not to do this yet (pursuant to Article 6.1(7)).

4. The certificate is signed by the chair of the Board of Examiners and the dean of the faculty.

5. The certificate includes a list of the examination components.

6. The Board of Examiners includes a diploma supplement as referred to in Article 7.11(4) of the Act with the certificate. This diploma supplement is based on the model adopted by UM’s Executive Board, which is in compliance with the agreed European standard format.

7. The Board of Examiners may award the ‘cum laude’ or ‘summa cum laude’ designation in accordance with the provisions in the Rules and Regulations.

8. Students who have passed more than one exam and who cannot be issued a certificate will upon request, receive a statement issued by the Board of Examiners which at least indicates the exams which they passed.
Article 6.4 Grade point average (GPA)
The diploma supplement referred to in Article 6.3(7) indicates the final grade point average (GPA) as specified in the Rules and Regulations, to provide a reflection of the student’s academic performance.

Article 6.5 Honours programme certificate
1. If one of the honours programmes referred to in Articles 3.11 or 3.12 is successfully completed, a separate certificate indicating this will be issued in addition to the one referred to in Article 6.3. The separate certificate is based on the model that UM’s Executive Board has adopted.
2. To receive this certificate, the student must, in addition to the requirements for the regular bachelor’s examination, satisfy the condition that the student has not been engaged in fraud with respect to an exam or a part thereof or has not been suspended due to other misconduct.
3. The Board of Examiners decides whether the student has satisfied all the specific requirements of the honours programme and can impose additional conditions in the Rules and Regulations.

Article 6.6 Right of appeal
Within six weeks after the decision by the examiner and the board of examiners is announced, the student may appeal this decision to UM’s Complaint Service Point. The appeal must be signed, must include a date and the name and address of the party lodging the appeal, must indicate the grounds for the appeal and, if possible, must include a copy of the decision being appealed.

Section 7 | Study Guidance and Advice

Article 7.1 Study progress administration
The faculty records the students’ individual study results and makes them available through the Student Portal.

Article 7.2 Study guidance
1. The faculty will provide for the introduction and study guidance for students registered for the programme, which also includes orientating them regarding possible study paths in and outside the study programme
2. The study guidance includes:
   a. an introduction during the first week of the first semester of the first academic year;
   b. assignment of a mentor for the first year in the bachelor’s programme;
   c. group and individual advice on possible study paths in and outside the programme, partly with a view to the professional options after the master’s programme and opportunities for immediately entering the labour market after obtaining the bachelor’s diploma;
   d. group and individual advice on study skills, study planning and choices of continuing study paths;
   e. offering referrals and help if the student experiences problems during the study;
   f. offering referrals for a more appropriate study programme to students who deregister for the programme before 1 February of the year of initial registration;
   g. giving a (negative) binding study advice (BSA) on the continuation of studies.
The Department of Data Science and Knowledge Engineering (DKE) focuses its attention to the bachelor’s programme in Data Science and Artificial Intelligence as well to the master’s in Artificial Intelligence and in Data Science for Decision Making.
Article 7.3 Study advice for the propaedeutic phase
1. At the end of the first year of registration for the propaedeutic phase of the bachelor’s programme, the (negative) binding study advice (BSA) is issued by the BSA Committee on behalf of the Dean to each student regarding continuation or not of the programme.
2. Subject to the provisions in the first paragraph, the BSA Committee on behalf of the Dean may issue the advice to the student as long as the student has not passed the exams related to the propaedeutic phase programme components.

Article 7.4 Standards
1. At the end of the first year of registration of the bachelor’s programme, the student must have earned at least 45 credits of the first course year.
2. When determining the number of credits obtained as referred to in paragraph 7.4.1 all credits obtained in the 1st year are included, except those for exemptions.

Article 7.5 Negative binding study advice (nBSA)
1. If the advice referred to in Article 7.3 is negative, the advice also entails a rejection for the bachelor’s programme.
2. The rejection referred to in paragraph 1 is binding and means that the student cannot register for the bachelor’s programme for the next 6 academic years.
3. The nBSA is issued to a student who was registered for the full-time programme at any time during the first academic year and who obtained less than 45 credits at the end of the academic year concerned.
4. Students who apply to deregister before 1 February of the first year of registration will not be issued with a study advice as referred to in Article 7.3.1. If a student re-registers in a subsequent academic year, the advice referred to in Article 7.3.1 will be issued in the next academic year based on the credits obtained in that academic year and in accordance with the regulations applicable to that year.
5. Before the nBSA is issued, the student will be given the opportunity to be heard by the BSA Committee.

Article 7.6 Procedure
1. No later than in the month of March of the first academic year, a student who has earned less than 24 credits at that time is given a written warning.
2. The warning indicates the period within which the study results must be improved.
3. In the month of August of the first academic year, students meriting negative binding study advice are given written notice that the BSA Committee intends to issue this advice to them.
4. Before the nBSA is issued, the student will be given the opportunity to be heard. The student will be informed of this in the notice referred to in paragraph 3. The hearings will take place in mid-August. A minimum of two members of the BSA Committee will attend the hearing.
5. The study adviser is informed which students merit negative binding study advice. The study adviser may, upon request or otherwise, make a recommendation to the BSA Committee.
6. After the student has been heard, the BSA Committee will determine whether to issue the nBSA to the student.
7. The student receives written notice of the nBSA decision by 31 August at the latest.
8. The written notice is signed by the Chair of the BSA Committee.
9. An appeal against the nBSA decision may be lodged with UM’s Complaint Service Point within six weeks of the date on which the decision was announced.
Article 7.7 Personal circumstances
1. In issuing the study advice, the BSA Committee takes the personal circumstances referred to in paragraph 2 of this Article into account.
2. Personal circumstances which may be considered include:
   a. Illness on the part of the student concerned;
   b. physical, sensory or other impairments which the student concerned has;
   c. pregnancy on the part of the student concerned;
   d. special family circumstances;
   e. administrative activities as referred to in Article 2.1(1) under (e), (f) and (g) of the Implementation Decree for the Act 2008 [Uitvoeringsbesluit WHW 2008];
   f. participation in top-level sport;
   g. circumstances other than those referred to in subparagraphs a. to f. which, if they were not to be honoured by the faculty board, would result in excessive unfairness.
3. To ensure that the student receives the best possible support, he/she must notify the study adviser of the personal circumstances as soon as possible.

Article 7.7b
1. If it is impossible to issue advice on the student’s suitability for the programme due to personal circumstances that occurred in the first year, contrary to Article 7.5 this advice may be postponed to a later moment during the propaedeutic phase. An adapted standard may be used for the student in question.
2. If the advice is postponed, a BSA will be issued by the end of the following year of registration in the programme. The postponed advice will be positive if the (adapted) BSA standard is met. The student will receive a negative BSA if he/she has not achieved the (adapted) BSA standard.
3. At the student’s request, the BSA Committee will take personal circumstances into account when deciding to issue a BSA. This decision will also be based on the student’s study behaviour, agreements and/or study plan made with the student adviser, the time at which the personal circumstances were reported and the study results achieved at the end of the first year of study.

Article 7.8 Hardship clause
In exceptional cases in which application of the study advice rules would lead to unreasonable treatment or serious unfairness, the faculty board can deviate from the stated regulations in the student’s favour.

Section 8 | Transitional and Final Provisions

Article 8.1 Amendments
1. Amendments to these regulations may be adopted in a separate decision by the faculty board, after a recommendation and consent from the programme committee and after consent from or consultation with the faculty council.
2. An amendment in these regulations will not pertain to the current academic year, unless the interests of the students will not reasonably be harmed as a result.
3. In addition, amendments may not affect, to the students’ detriment, a decision regarding a student which has been taken by the Board of Examiners pursuant to these regulations.

Article 8.2 Notice
1. The faculty board ensures that proper notice is given of these regulations, the rules
and regulations adopted by the board of examiners, and any changes to these
documents, by, for example, placing such notice on the faculty website and/or the
Student Portal.
2. Any interested party may obtain a copy of the documents referred to in the first
paragraph from the Student Affairs office.

Article 8.2a Evaluation
The Faculty Board will ensure that the education of the programme is regularly
evaluated, assessing at least – for the purpose of monitoring and if necessary adapting
the student workload – the amount of time students need to complete their duties as
set out therein.

Article 8.3 Unforeseen cases/safety net scheme
1. In cases not covered or not clearly covered by these regulations, decisions are taken by
or on behalf of the faculty board, after it has consulted with the Board of Examiners.
2. In individual cases in which application of the Education and Examination Regulations,
except for the study advice rules, would lead to manifestly unreasonable results, the
Board of Examiners can deviate from the stated regulations in the student’s favour.

Article 8.4 Effective date
This Regulation will come into force on 1 September 2020 and will apply
for the academic year 2020/2021.

Adopted by the faculty board on May 19, 2020.
Appendix 1 | Blocks with limited capacity (see Article 4.5)

<table>
<thead>
<tr>
<th>Block</th>
<th>Faculty</th>
<th>Number of participants</th>
<th>Offered in semester</th>
</tr>
</thead>
<tbody>
<tr>
<td>2.1-3.6 FSE/DKE</td>
<td>t.b.a. - KE@Work</td>
<td>-</td>
<td>Year 2 and 3</td>
</tr>
<tr>
<td>2.1-3.6 FSE/DKE</td>
<td>t.b.a. - MaRBLe 2.0</td>
<td>-</td>
<td>Year 2 and 3</td>
</tr>
<tr>
<td>3.1-3.3 FSE/DKE</td>
<td>t.b.a. - Exchange programme</td>
<td>-</td>
<td>Semester 1 of year 3</td>
</tr>
<tr>
<td>3.1-3.3 FSE/DKE</td>
<td>t.b.a. - Minor Entrepreneurship</td>
<td>-</td>
<td>Semester 1 of year 3</td>
</tr>
<tr>
<td>3.1-3.6 FSE/DKE</td>
<td>t.b.a. - Educational Minor</td>
<td>-</td>
<td>Year 3</td>
</tr>
</tbody>
</table>

Appendix 2 | Equivalent pre-university education

A student that holds a non-Dutch diploma that is equivalent to the Dutch VWO diploma is admissible to the bachelor programme in Data Science and Knowledge Engineering in the following cases:

- Belgian diplomas: ASO.
- German diplomas: Abiturzeugnis or an equivalent pre-university high-school diploma with a sufficient amount of Math education.
- Other countries: A pre-university high-school diploma equivalent to the Dutch VWO diploma with a sufficient amount of math education (see list on the DKE website).

If a student’s diploma is not included in the list on the website, it must be evaluated first.

Appendix 3 | Regulations for exchange programmes to study abroad

1. Components attained in connection with an exchange programme adopted by the faculty board may be incorporated into the examination for the programme for which the student is registered.
2. In principle, a student pursuing part of his/her studies in a foreign country does this as part of the DKE’s exchange programme at one of its exchange partners.
3. Studies in connection with the exchange programme constitute part of the studies at UM.
4. The certificate may not be issued until the exchange programme has been completed and the results in the foreign country have been incorporated. The results will be taken into account for classification on the same basis as the results attained in the faculty.
5. In connection with the requirements regarding the level and content of the examination for the programme concerned, several conditions apply to incorporation. The Board of Examiners assesses whether a component may be incorporated into the examination. Incorporation is only possible with the Board of Examiners’ permission. The following conditions must be satisfied:
   a. the component is academic in nature;
   b. in terms of content, the component does not overlap with other examination components (Article 3.6 of the Education and Examination Regulations for Bachelor’s Programmes);
   c. the component was attained in connection with an exchange programme, except for the situation described below;
   d. the student requested written permission for incorporation beforehand, in principle, by submitting a request with appendices to the Student Affairs Office, which forwarded the information to the Board of Examiners;
   e. the student will submit additional information to the Board of Examiners upon request.
6. Point 2 may be deviated from if the student has good substantive reasons for studying in a foreign country at a university, which is not an exchange partner. Whether a foreign university is an exchange partner during the current year may be ascertained in the student handbook. The following procedure applies in such a situation:

- The student must timely submit (by letter or e-mail) a written request, stating reasons, to the Study Adviser. In addition to the student’s personal and study progress information, the request must include information about the university where the student wishes to study, as well as about the student’s substantive reasons for the request. These reasons must relate to the student’s own programme.
- The request is assessed by the Director of Studies. In this assessment, the level and the available courses of the university concerned are compared to the exchange partners; the Director of Studies does not provide a judgment about the student’s specific choice of subject is not be provided, but only about general questions regarding the level and so forth.
- If the judgment about the university (regarding its level or otherwise) is favourable, the Study Adviser will notify the student and the Board of Examiners. The student will then follow the normal exchange programme procedure and will also follow the normal approval procedure with the Board of Examiners for incorporation of subjects.

All costs involved for studying at a non-exchange partner are on behalf of the student.

7. The following rules apply to incorporation by the Board of Examiners of results attained in foreign countries:

a. results must be provided to the Board of Examiners through clear, original written notices by the university concerned. In addition to personal information about the student and information about the university, these notices should preferably include the following information: subject code and name; subject level; subject study load; and exam date and result;

b. incorporation will only be possible if there has been a sufficient assessment. A mere statement that the subject was ‘taken’ or words to that effect will not be enough. A result which constitutes a pass in the country concerned will also be considered a pass here, and a fail in a foreign country will also be considered a fail here;

c. the study load conferred by the foreign university will be followed here without adjustment. If, for example, 5 ECTS are conferred by the university concerned for the subject, the subject will also count for 5 ECTS at the faculty, regardless of how difficult or easy the subject was for the student.

d. if the study load in the foreign country is not expressed in ECTS units but in other units, the Board of Examiners will convert the study load into ECTS units. The conversion will be based on the study load for a full year of study or a full programme in relation to a study load of 60 ECTS for a year at UM or 180 ECTS for a three-year bachelor’s programme.

8. Study delay caused by the study abroad is the responsibility of the student.

9. Insofar as the Education and Examination Regulations and this appendix do not address matters relating to incorporation of components attained elsewhere, the Board of Examiners will decide on the matter.
Appendix 4 | Accountability for the language of instruction

The choice for the language of instruction of the programme is in line with the UM Code of Conduct on language in accordance with the Dutch Higher Education and Research Act (WHW) art. 7.2.

Because of the specific educational nature and profile of the programme, teaching and examinations are conducted in English. This guarantees the quality of education, because:

• The content of the programme has an international orientation and focus. Language of main course materials (handbooks, papers, tools) discussing or supporting the latest advances in AI or Data Science are in English.
• The academic community is internationally oriented and the staff is international. Project-Centred Learning implements the international classroom concept, which requires a common international language. Students participate in international collaboration by exchange programmes to study abroad, or conduct their thesis at companies/institutes where English is the lingua franca. There is a staff exchange with RWTH Aachen, such that the courses on Parallel Programming and Large Scale IT and Cloud Computing in the programme are coordinated by RWTH Aachen.
• The labour market demand is internationally oriented (English speaking). Alumni typically end up in jobs at companies/institutes where English is the main language.
• The student intake and current population is internationally diverse and English is the common language (26% Dutch; 74% non-Dutch in 2018-2019).
5.2 Master’s programmes Artificial Intelligence and Data Science for Decision Making

Section 1 | General provisions
Article 1.1 Applicability of the regulations
Article 1.2 Definitions

Section 2 | Admission Master Programme
Article 2.1 Admission
Article 2.2 Eligibility for admission
Article 2.3 Certificate of admission
Article 2.4 Language requirement with non-Dutch diplomas
Article 2.5 Board of Admission
Article 2.6 Admission test dates
Article 2.7 Bridging Programme

Section 3 | Content and Structure of the Programme
Article 3.1 Programmes
Article 3.2a Objectives of the programme Artificial Intelligence
Article 3.2b Objectives of the programme Data Science for Decision Making
Article 3.2c Qualifications of the programme Artificial Intelligence
Article 3.2d Qualifications of the programme Data Science for Decision Making
Article 3.3 Form of the programme
Article 3.4 Language of instruction
Article 3.5 Communications and announcement of decisions
Article 3.6 Study load
Article 3.7 Content of the programme
Article 3.8 Electives
Article 3.9 Flexible programme and flexible master’s
Article 3.11 Study Abroad
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Section 4 | Education
Article 4.1 Courses; composition; actual design
Article 4.2 Prior knowledge
Article 4.3 Course registration
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Section 5 | Assessment
Article 5.1 General
Article 5.2 Marks
Article 5.3 Order of exams
Article 5.4 Scheduling and frequency of the exams
Article 5.5 Registration for exams
Article 5.6 Form of the exams
Article 5.7 Oral exams
Article 5.8 Assessments in exceptional cases
These regulations apply to the education and exams and examinations of each of the two master’s programmes Artificial Intelligence and Data Science for Decision Making, each referred to as ‘the programme’ in the remainder of this document. The programme is offered by the Faculty of Science and Engineering, hereafter called the faculty, at the Department of Data Science & Knowledge Engineering, hereafter to be called the department.

The regulations were adopted by the Faculty Board and the School Board of the School of Information Technology, after advice and consent of the programme committee and after the consent from or in consultation with the faculty council. The regulations will take effect on 1 September 2020 for the 2020-2021 academic year. These regulations also apply to students from other programmes, faculties or
institutions of higher education, insofar as they follow components of the programme to which these Education and Examination Regulations apply. For components of the programme that students follow at another degree programme, faculty or institution of higher education, the Education and Examination Regulations for the other programme, faculty or institution apply to the component in question.

Article 1.2 Definitions
In these regulations, the following definitions apply:

a. The Act: the Higher Education and Scientific Research Act (Wet op het hoger onderwijs en wetenschappelijk onderzoek);
b. Student: a person who is registered at the university for education and/or to take exams and the examination of the programme;
c. Course: a study unit of the programme within the meaning of the Act;
d. Course year: year 1 or year 2 of the programme;
e. Academic year: the period from 1 September of a calendar year up to and including 31 August of the following calendar year;
f. Programme: each of the master’s programmes referred to in Article 1.1 of these regulations, consisting of a coherent whole of study units;
g. Exam: a component of the examination as referred to in Article 7.10 of the Act;
h. Practical: skill exercise as referred to in Article 7.13(2)(d) of the Act, in one of the following forms:
  - writing a thesis;
  - carrying out a (group) project;
  - performing a research assignment;
  - developing a software program;
  - writing a paper, creating a technological design or performing another written assignment;
  - participating in field work or a field trip;
  - completing an internship;
  - participating in an activity intended to develop certain skills
i. Project Skill: practical training associated as part of the project.
j. Examination: the final examination for the master’s programme;
k. Credit: a unit expressed in ECTS credits, with one credit equalling 28 hours of study;
l. Board of Examiners: the board referred to in Article 7.12 of the Act;
m. Board of Admission: the board responsible for judging the admissibility of the candidate to the programme;
n. Programme Committee: the representation and advisory body that carries out the duties described in Article 9.18 and 9.38c of the Act.
o. Examiner: the person designated by the Board of Examiners to administer exams and to determine the results of such exams;
p. Faculty Board: the faculty board referred to in Article 9.12 of the Act;
q. Semester: part of an academic year, either starting first of September and running for 20 educational weeks, or starting first of February running for 21 educational weeks.
r. Block: part of a semester during which educational activities take place;
s. AI: Artificial Intelligence;
t. DSDM: Data Science for Decision Making;
u. KE: Data Science & Knowledge Engineering;
v. DKE: Department of Data Science & Knowledge Engineering;
Section 2 | Admission Master Programmes

Article 2.1 Admission
Candidates who have obtained either:
• the Bachelor of Science degree in Data Science and Knowledge Engineering at the UM, or
• the Bachelor of Science degree in Computer Science at Hasselt University,
• the Bachelor of Science degree in scientific programming at Aachen University are eligible for admission to the programme.

Article 2.2 Eligibility for admission
Those students that did not obtain the degree referred to in Article 2.1, but who satisfy the requirements referred to in Article 2.3 and have received a certificate of admission, are eligible for admission to the programme.

Article 2.3 Certificate of admission
The certificate of admission referred to in Article 2.2 will be issued if the candidate concerned satisfies the following admission requirements;

Degree
1. A Bachelor of Science degree in Data Science and Knowledge Engineering or an equivalent diploma in a related field to Data Science and Knowledge Engineering.
2. Students having a professional Bachelor of Science degree in Data Science and Knowledge Engineering or a related field from a Dutch HBO study or equivalent, and successfully completed a tailored bridging programme as referred to in Article 2.7.

Additional requirements
• Candidates must provide a curriculum vitae and write a motivation essay of 2 pages in A4 format.
• Candidates with a non-EU/EEA diploma should show proof of English proficiency referred to in Article 2.4.
• The Board of Admission can additionally request as proof of analytical writing and quantitative reasoning abilities, a satisfactory Graduate Record Examination (GRE) score. A satisfactory GRE score implies a score of 4 or higher on the analytical writing section and 80% in the quantitative reasoning section (In case of a lower score, the Board of Admission can still decide that a candidate is admissible).

Article 2.3a Certificate of admission
The Board of Admission can approve specific bachelor programmes at Applied universities, for which students that completed specific components, consisting of a bridging minor in collaboration with DKE, are admissible to the Master programme.
Conditions are subject to approval by the Board of Admission and the bridging minor subject to approval by the Board of Examiners.

**Article 2.4 Language requirement with non-Dutch diplomas**

a. Holders of a non-Dutch diploma can only register if they have met the minimum English language requirement corresponding to IELTS (International English Language Testing System) with a score of at least 6.5.

b. The requirement referred to under (a) is met if the person concerned has obtained one of the following diplomas or certificates:

- a completed bachelor’s or master’s study programme where the language of instruction is English;
- an International or European Baccalaureate, a US high school diploma or UK GCE A-levels, or
- can demonstrate sufficient proficiency in English, for example through English taught courses, internships or work experience in an English environment, or
- can submit one of the following language test certificates.
  - IELTS (6.5 minimum)
  - TOEFL Paper-based test (575 minimum)
  - TOEFL Internet test (90 minimum)
  - TOEFL Computer test (232 minimum)
  - TOEIC (720 minimum) the modules ‘listening and reading’ and ‘speaking and writing’ must be completed successfully
  - Cambridge [Advanced (CAE) Grade C, First Certificate in English (FCE) Grade A, First Certificate in English (FCE) Grade B or
  - similar accredited certification

**Article 2.5 Board of Admission**

1. The Board of Admission for the programmes is responsible for assessing eligibility for admission and issuing the certificate of admission to the programme. The Board of Admission consists of three members appointed from the academic staff responsible for the programme curriculum, under which one chair.

2. The student adviser for the programme concerned is appointed as an adviser, also the Secretary.

3. The Faculty Board appoints the members after consulting with the Programme Committee.

**Article 2.6 Admissions test dates**

1. The admissions test takes place twice a year.

2. A request for admission to a programme must be submitted to the Board of Admission before 1 May, for a non-EU request, or before 1 July, for an EU request.

3. In special cases, the Board of Admission can take up a request submitted after the closing date referred to in paragraph 2.

4. The Board of Admission decides on the request as soon as possible.

5. The candidate will be admitted subject to the condition that, by the relevant start date for the programme, he/she will have satisfied the requirements stated in Article 2.3 and 2.4 regarding knowledge, understanding and skills, as evidenced by the certificates for the programmes taken by him/her.

**Article 2.7 Bridging Programme**

1. Students having a professional Bachelor of Science degree in Data Science & Knowledge Engineering or a related field (e.g., Mathematics, Computer Science or
Artificial Intelligence) from a Dutch HBO study or equivalent have the possibility to meet the requirements as referred to in Article 2.3 by successfully completing a tailored bridging programme.

2. The bridging is composed of key courses taken from the second and third year of the Bachelor of Science in Data Science and Knowledge Engineering together worth 30 ECTS and decided upon by the Board of Examiners after advice of the Board of Admission.

3. The bridging programme must be successfully completed within 12 months. After this term has expired or in case of unsatisfactory participation either in a qualitative or quantitative way, he/she may be debarred from further participation.

Section 3 | Content and Structure of the Programme

Article 3.1 Programmes
The master’s programme is one of the two following programmes:
1. the Artificial Intelligence programme;
2. the Data Science for Decision Making programme;

Article 3.2a Objectives of the programme Artificial Intelligence
The goals of the Master’s programme Artificial Intelligence are as follows:
1. A university education within the framework of Project-Centred Learning (PCL) and the UM teaching philosophy.
2. Added depth to the student’s knowledge in the chosen academic field.
3. The opportunity to broaden his/her education into other disciplines.
4. Specialised knowledge, skills and understanding in the field of Artificial Intelligence.
5. A preparation for research programmes in the field of Artificial Intelligence.

The programme includes sufficient aspects furthering the student’s university education, in particular regarding:
• independent academic thinking and conduct;
• academic communication in the English language; and
• application of the knowledge gained in the discipline in a broader societal context.
Specifically, the Master’s programme in Artificial Intelligence provides a thorough knowledge of algorithms, methods, and techniques from the fields of artificial intelligence and data science, such as agent technology, search techniques, machine learning, information retrieval & text mining, and computer games to model, analyse and implement intelligent software in a variety of contexts.

Article 3.2b Objectives of the programme Data Science for Decision Making
The goals of the Master’s programme Data Science for Decision Making at DKE are as follows:
1. A university education within the framework of Project-Centred Learning (PCL) and the UM teaching philosophy.
2. Added depth to the student’s knowledge in the chosen academic field.
3. The opportunity to broaden his/her education into other disciplines.
4. Specialised knowledge, skills and understanding in the field of Data Science, Applied Mathematics and Operations Research.

The programme includes sufficient aspects furthering the student’s university education, in particular regarding:
• independent academic thinking and conduct;
• academic communication in the English language; and
• application of the knowledge gained in the discipline in a broader societal context.

Specifically, the Master’s programme in Data Science for Decision Making provides a thorough knowledge of relevant algorithms, methods, and techniques from the fields of Data Science, Applied Mathematics, Operations Research, Artificial Intelligence and Machine Learning to model, analyse, optimise and implement (intelligent) software in a variety of contexts.

Article 3.2c Qualifications of the programme Artificial Intelligence
The qualifications of the programme Artificial Intelligence are as follows:

I. Knowledge and understanding
The student/graduate demonstrates knowledge and understanding in a field of study that builds upon and supersedes the Bachelor degree. Knowledge, understanding, and abilities are typically at a level at which the student/graduate is able to formulate a feasible research plan in one’s own specialisation.

Qualifications:
1. Advanced understanding of key areas in Artificial Intelligence, in particular in the subfields of machine learning, agent technology and search techniques.
2. Specialist knowledge of at least one of the key areas in Artificial Intelligence, up to a level that the Master can appreciate the forefront of research in that field.

II. Applying knowledge and understanding
Students/graduates can apply their knowledge and understanding in a manner that shows a scientific approach to their work or vocation. They are able to handle complex and ill-defined problems for which it is not a priori known if there is an appropriate solution, they know how to acquire the necessary information to solve the sub-problems involved, and they know how to proceed with problems for which there is no standard or reliable route to the solution.

Qualifications:
3. The ability to formulate a project plan for an open problem in a field related to Artificial Intelligence in general and to one’s own specialisation in particular.
4. The ability to judge the feasibility of a proposal to lead to a solution or design as specified.
5. The ability to contribute autonomously and with minimal supervision to an interdisciplinary project team and to profit from the abilities, the knowledge, and the contributions of other team members.
6. The ability to choose, apply, formulate, and validate models, theories, hypotheses, and ideas from the key areas of Artificial Intelligence.
7. The ability to submit an argument in the exact sciences (or humanities) to critical appraisal and to incorporate its essence in the solution of Artificial Intelligence problems.
8. The ability to translate academic knowledge and expertise into social, professional, economic, and ethical contexts.
9. Awareness of, and responsibility concerning, the ethical, normative and social consequences of developments in science and technology, particularly resulting from original contributions.

III. Making judgements
The student/graduate is able to formulate an opinion or course of action on the basis of incomplete, limited and in part unreliable information.
Qualifications:

10. Competence in the search and critical processing of all sources of information that helps to solve an open and ill-defined problem.

11. The ability to demonstrate a professional attitude conform the (international) scientific conduct in Artificial Intelligence.

12. The ability to provide and receive academic criticism conform the standards in one’s own specialism of Artificial Intelligence research.

13. The ability to formulate an opinion and to make judgements that include social and ethical responsibilities related to the application of one’s own contributions.

14. The ability to be able to judge the quality of his or her work or the work of others from the scientific literature.

IV. Communication

The student/graduate can communicate information, ideas, problems and solutions to audiences of specialists in (other) research areas and to a general audience.

Qualifications:

15. The student/graduate has academically appropriate communicative skills; s/he can:
   1. Communicate original ideas effectively in written form,
   2. Make effective oral presentations, both formally and informally, to a wide range of audiences
   3. Understand and offer constructive critiques of the presentations of others.

V. Learning skills

The student/graduate has developed those learning skills that are necessary for a successful further career at the highest professional level. The Master is able to detect missing knowledge and abilities and to deal with them appropriately.

Qualifications:

16. Being able to reflect upon one’s competences and knowledge and, if necessary, being able to take the appropriate corrective action.

17. The ability to follow current (scientific) developments related to the professional environment.

18. Showing an active attitude towards continued learning throughout a professional career.

Article 3.2d Qualifications of the programme Data Science for Decision Making

The qualifications of the programme Data Science for Decision Making are as follows:

I. Knowledge and understanding

The student/graduate demonstrates knowledge and understanding in a field of study that builds upon and supersedes the Bachelor degree. Knowledge, understanding, and abilities are typically at a level at which the Master is able to formulate a feasible research plan in one’s own specialization.

Qualifications:

1. Advanced understanding of key areas in Applied Mathematics and Data Science, in particular in the subfields Pattern Recognition, Operations Research, Machine Learning, and Intelligent Decision Making.

2. Specialist knowledge of at least one of the key areas in Data Science and Decision Making, up to a level that the Master can appreciate the forefront of research in that field.
II. Applying knowledge and understanding
Students/graduates can apply their knowledge and understanding in a manner that shows a scientific approach to their work or vocation. They are able to handle complex and ill-defined problems for which it is not a priori known if there is an appropriate solution, they know how to acquire the necessary information to solve the sub-problems involved, and they know how to proceed with problems for which there is no standard or reliable route to the solution.

Qualifications:
3. The ability to formulate a project plan for an open problem in a field related to Applied Mathematics and Data Science (in general, and to one’s own specialization in particular.
4. The ability to judge the feasibility of a proposal, taking into account the availability of data, to lead to a solution or design as specified.
5. The ability to contribute autonomously and with minimal supervision to an interdisciplinary project team and to profit from the abilities, the knowledge, and the contributions of other team members.
6. The ability to choose, apply, formulate, and validate models, theories, hypotheses, and ideas from the key areas of Applied Mathematics and Operations Research.
7. The ability to transform, analyse and interpret data and to extract information from it, using techniques from Data Science.
8. The ability to submit an argument in the Exact Sciences (or Humanities) to critical appraisal and to incorporate its essence in the solution of problems in Applied Mathematics and Data Science.
9. The ability to translate academic knowledge and expertise into social, professional, economic, and ethical contexts.
10. Awareness of, and responsibility concerning, the ethical, normative and social consequences of developments in science and technology, particularly resulting from original contributions.

III. Making judgements
The student/graduate is able to formulate an opinion or course of action on the basis of incomplete, limited and in part unreliable information.

Qualifications:
11. Competence in the search and critical processing of all sources of information that helps to solve an open and ill-defined problem.
12. The ability to demonstrate a professional attitude conform the (international) scientific conduct in Data Science for Decision Making.
13. The ability to provide and receive academic criticism conform the standards in one’s own specialism of Applied Mathematical and Data Science research.
14. The ability to formulate an opinion and to make judgements that include social and ethical responsibilities related to the application of one’s own contributions.
15. The ability to judge the quality of his or her work, or the work of others from the scientific literature.

IV. Communication
The student/graduate can communicate information, ideas, problems and solutions to audiences of specialists in (other) research areas and to a general audience.

Qualifications:
16. The Master has academically appropriate communicative skills; s/he can:
   a. Communicate original ideas effectively in written form,
   b. Make effective oral presentations, both formally and informally, to a wide range of audiences
c. Understand and offer constructive critiques of the presentations of others.

V. Learning skills
The student/graduate has developed those learning skills that are necessary for a successful further career at the highest professional level. The Master is able to detect missing knowledge and abilities and to deal with them appropriately.

Qualifications:
17. Being able to reflect upon one’s competences and knowledge and, if necessary, being able to take the appropriate corrective action.
18. The ability to follow current (scientific) developments related to the professional environment.
19. Showing an active attitude towards continued learning throughout a professional career.

Article 3.3 Form of the programme
This is a full-time programme. The programme commences twice a year, in September and February of the following calendar year.

Article 3.4 Language of instruction
1. The programme is given in English (accountability for this language of instruction is given in Appendix 5). Components of the programme may be in Dutch or in another common language in the EU.
2. Dutch, English, French and/or German texts may be used in the education and exams in the programme.

Article 3.5 Communications and announcement of decisions
1. The Faculty board, the Board of Examiners, the DKE Management Team and the examiners may use the Student Portal and email via the UM account for communications relating to the education and examinations.
2. The Faculty Board, the Board of Examiners, the DKE Management Team and the examiners may use the Student Portal and email through the UM account to announce decisions.
3. The student must regularly check his/her university e-mail address, the Faculty website and the digital learning environment. Information disseminated via e-mail, the digital learning environment or the website will be assumed to be known.

Article 3.6 Study load
The programme has a study load of 120 credits (ECTS), with one credit equalling 28 hours of study.

Article 3.7 Content of the Programmes
The components of the Artificial Intelligence programme are listed in Appendix 1a. The components of the Data Science for Decision Making programme are listed in Appendix 1b. The student’s choice of electives is subject to the Board of Examiners’ approval.

Article 3.8 Electives
The first semester of year 2 of the programme contains electives:
1. The student selects one or more components with a total study load equal to 30 ECTS.
2. The components mentioned in Appendix 1, may be chosen as electives.
3. The student may - subject to prior approval by the Board of Examiners of the other programme - choose to take components given by another UM department/faculty, another Dutch university or a foreign university.

4. The Board of Examiners may withhold the approval referred to in paragraph 3. A particular reason for withholding is if, in its judgement, the proposed elective is in terms of content similar to components taken previously by the student and would result in duplication. If components of the electives overlap in whole or in part, the Board of Examiners can decide to limit the contribution of these components towards the overall assessment by deducting of ECTS in proportion to the overlap.

5. To attain the certificate for the examination for the master’s programme, the student must have obtained at least 78 ECTS of the educational programme through components provided by DKE.

6. For courses passed at a higher education institute prior to the start of the master’s programme, an exemption may only be granted on the basis of Article 5.15.

Article 3.9 Flexible programme and flexible master’s
1. A student registered for one of the faculty’s programmes may, under certain conditions, formulate an educational programme of his/her own which is different from the educational programme stated in Article 3.7. The composition of such a programme must be approved beforehand by the Board of Examiners.

2. The flexible programme must have a study load of 120 credits.

Article 3.10
Article cancelled

Article 3.11 Study Abroad
1. Students can apply to study abroad for a semester, at another University with whom Maastricht University has an Agreement of Exchange.

2. Nomination is decided on by the Board of Examiners based on study progress as mentioned in Article 5.3.1 and motivation of the student.

3. This study abroad will take place in Semester 1 of year 2 and has a study load of 30 ECTS.

4. The selected course programme has to be approved by the Board of Examiners.

5. Further regulations can be found in Appendix 4.

Article 3.12 The examination
The examination for the master’s programme Artificial Intelligence consists of the following components:

a. Course year 1, offered from September 2011 or later:
exams in blocks 1.1, 1.2, 1.4, 1.5, listed in Appendix 1a;
research projects in blocks 1.3 and 1.6.

b. Course year 2, offered September 2011 or later:
internship, or research project, or electives, with a total of 30 ECTS;
master thesis, with a total of 30 ECTS

The examination for the master’s programme Data Science for Decision Making consists of the following components:

a. Course year 1, offered September 2011 or later:
exams in blocks 1.1, 1.2, 1.4, 1.5, listed in Appendix 1b;
research projects in blocks 1.3 and 1.6.

b. Course year 2, started September 2011 or later:
internship, or research project, or electives, with a total of 30 ECTS; master thesis, with a total of 30 ECTS.

Section 4 | Education

Article 4.1 Courses; composition; actual design
1. For the programme components, courses are given with the study load stated in Article 3.6.
2. The education is given in the form of classes, project groups, practical training, lectures, individual supervision, or otherwise. On average, the student has 13 hours of face-to-face time per week, but this can differ per period. For further details, please see the student handbook.
3. The educational programme includes 41 weeks per year. During this period students are expected to be available for educational activities.

Article 4.2 Prior knowledge
The desired prior knowledge to successfully participate in each course is indicated in the student handbook.

Article 4.3 Course registration
The student may participate in a course after he/she has timely registered for it through the Student Portal.

Article 4.4 Attendance and best-efforts obligation
1. Project skill trainings and project meetings are mandatory. In addition, each student is expected to participate actively in doing tasks with respect to the project and to cooperate actively with their group in order to successfully finish the project.
2. Students who have not met the requirements in paragraph 1 and/or students whose absence or inactivity during the project has been marked as inexcusable by the project coordinator and/or students that have a substandard contribution to the group work will not receive a pass for the project concerned.
3. Attendance and participation in other education activities may be part of an exam when announced in the student handbook or student portal and prior approval of the Board of Examiners.
4. For resits that are offered for practical assignments, the examiner can require that the student has made a reasonable attempt at the regular opportunity for that assignment. This requirement must be communicated upfront to the students on student portal. The Board of Examiners can impose further restrictions in the Rules and Regulations.

Article 4.5 Participation
The blocks mentioned in Appendix 2 are available to a limited number of students enrolled at UM, provided that the students enrolled in the programme will anyhow be placed in the blocks belonging to the compulsory part of their programme, in compliance with the periods the faculty designated to them. For blocks belonging to the optional parts of the master’s programme, most places will be reserved for students enrolled in the programme.
Article 4.6 Practical training
Some courses indicated in Appendix 1 include practical training for which additional regulations apply as specified in the Rules and Regulations.

Section 5 | Assessment

Article 5.1 General
1. During a course, the student will be tested for academic training and the extent to which the student has sufficiently achieved the stated learning objectives.
2. The student handbook describes the achievements the students must make to pass the course and the criteria on which the student is assessed. Any amendments are published on student portal.
3. The Rules and Regulations describe the assessment procedure.

Article 5.2 Marks
1. Marks are awarded on a scale of 1 to 10. Whole grades are awarded to exams. Both half and whole grades can be awarded for study components of at least 10 ECTS as well as for projects
2. The student must receive a final mark of 6.0 or higher to pass the component.
3. The highest result attained determines the final mark.
4. NG (no grade) can be assigned as a result of plagiarism or academic dishonesty; or when assessment is incomplete and no grade can be assigned. An NG automatically constitutes a fail and no credit is awarded.

Article 5.3 Order of exams
1. If the student obtained at least 40 ECTS of course year 1, he or she can sit exams of course year 2.
2. If the student has obtained at least 40 credits in course year 1 and at least 60 ECTS overall, he/she may start the master thesis in course year 2.
3. In conformance with article 7.30 paragraph 3 of the Act, the Board of Examiners may grant a student permission to sit other exams than referred to in paragraphs 1 and 2.
4. If a student deviates from the sequence as described in paragraphs 1 and 2, without permission from the Board of Examiners, the result of the component in question can be declared invalid.

Article 5.4 Scheduling and frequency of the exams
1. Students can take exams twice per academic year on dates to be determined by the Board of Examiners: once during or directly after the block (first sit for the exam) and once before the end of the semester in question (resit option).
2. Once a student successfully passes an exam, he/she can resit the exam upon permission from the Board of Examiners.
3. In exceptional cases, the Board of Examiners can decide that an exam may be taken at another time than determined in accordance with the first paragraph.

Article 5.5 Registration for exams
1. The student may take an exam for a course for which he/she is registered, after he/she has timely registered for the exam through the Student Portal.
2. If a student deviates from the conditions as described in paragraph 1, the result of the component in question will be declared invalid.
Article 5.6 Form of the exams
1. In principle, the exams are written. ‘Written exams’ also include taking exams by computer.
2. In principle all exams and assignments with the exception of research projects are on an individual basis, unless explicitly announced otherwise.
3. Oral exams can only take place upon prior approval by the Board of Examiners.
4. The form of the exam is announced by the examiner at the start of the block.
5. Upon request, students with a disability may take exams in a manner, which accommodates their specific disability as much as possible. If necessary, the Board of Examiners will obtain expert advice from the faculty’s student advisor and/or the student dean at the Student Service Centre (SSC) before taking a decision in such matters.

Article 5.7 Oral exams
1. Oral exams are taken only by one person at a time, unless the Board of Examiners decides otherwise.
2. An oral exam is given by the examiner in the presence of second examiner, unless the Board of Examiners has decided otherwise.
3. Oral exams take place in public, unless the Board of Examiners or the relevant examiner decides otherwise in a special case.

Article 5.8 Assessments in exceptional cases
1. A participant can submit a request to the Board of Examiners for an individual assessment. This request may be granted if the participant has not passed the exam in question due to exceptional circumstances and not granting an individual assessment would result in an unacceptable study delay.
2. The following criteria apply to the granting of an individual assessment for the final component of the programme:
   - It must be the final study result to be obtained.
   - The study delay in case the individual assessment is not granted must be at least one semester.
   - The participant must have taken part in the last two regular exam opportunities for the exam for which he/she is requesting another assessment.

Article 5.9 Practicals and written assignments
1. The Board of Examiners may draw up guidelines for the practicals which include group projects and theses. The guidelines will be included in the Rules and Regulations.
2. The master’s thesis project will be evaluated by at least two examiners (the relevant supervisor and a second evaluator), at least two of them are affiliated with the programme for which the student is registered.

Article 5.10 Internships
1. DKE can offer internship vacancies or the student may find a relevant internship and DKE supervisor him/herself. For each internship, an internship proposal must be send to the Board of Examiners for approval.
2. The Board of Examiners appoints an examiner for the internship.
3. The student may undertake an internship supervised by DKE at most once during the programme.
4. The Board of Examiners may formulate guidelines for internships. The guidelines will be included in the Rules and Regulations.
Article 5.11 Determination and announcement of exam result
1. The Board of Examiners determines the standards for assessing each examination component. The standards are included in the Rules and Regulations.
2. The examiner determines the result of a written exam within 15 working days of the date on which it was taken and at least 10 working days before the resit, and provides the Student Affairs Office with the necessary information to apprise the student of the result.
3. The examiner determines the result of an oral exam within 24 hours and issues the relevant certificate to the student. If more than one student takes the same exam after each other, this period may be extended by up to five working days.
4. When the result of a written exam is announced, it will be indicated how the student can inspect the exam and file an appeal as referred to in Article 6.5.

Article 5.12 Right of inspection
1. Within 10 working days of the date on which the result of a written exam, including a computer-based exam, is announced, students may inspect their evaluated work.
2. Within the period referred to in paragraph 1, any interested party may, upon request, inspect the questions and assignments for the written exam and, if possible, the standards based on which it was assessed.

Article 5.13 Period of validity
1. Exams which have been passed are valid for an unlimited period. Contrary to the above the Board of Examiners may require the participant to take an additional or replacement exam or examination component for an exam which was passed more than six years ago if the student’s knowledge or insight that was examined is demonstrably outdated or the skills that were examined are demonstrably outdated.
2. If exceptional circumstances apply as referred to in Article 7.51 paragraph two of the Act, the period of six years in paragraph one will be extended by the duration of the financial support a student receives from the profiling fund.
3. Sub-tests and assignments that were passed within a component, which was not passed, will lose their validity after the academic year in which they were passed unless the Board of Examiners states otherwise.

Article 5.14 Retention period for exams
1. The exercises, answers and the evaluated work of the written tests will be retained in paper or digital form for two years after the exam/examination result is determined.
2. The master thesis and its evaluation will be kept for at least seven years after the evaluation.

Article 5.15 Exemption
1. The Board of Examiners may, at a student’s request and having heard the relevant examiner, grant the student an exemption from taking an exam if he/she demonstrates that he/she previously:
   i. either passed an exam for a university or higher professional education programme which was similar in terms of content and level or
   ii. gained sufficient knowledge and skills relevant to the exam concerned, either through work or professional experience.
2. An exemption may only pertain to an entire course and not a part thereof.
3. At most 40 credits for the programme may be earned based on the exemptions granted.
4. The master’s thesis is excluded from this exemption option.
5. The Board of Examiners will not grant any exemption based on exams passed by a student outside the programme during the period in which the student was barred by the Board of Examiners from taking exams for the programme because of fraud.
6. The same period of validity applies to exemptions as to exam results.

Article 5.16 Fraud
1. ‘Fraud’, including ‘plagiarism’, means actions or omissions by a student which make it impossible in whole or in part to properly evaluate his/her knowledge, understanding and skills. ‘Plagiarism’ means the presentation of ideas or words from one’s own or someone else’s sources without proper acknowledgment of the sources.
2. If the Board of Examiners determines that a student has engaged in fraud with respect to an exam or a part thereof, the Board of Examiners can take appropriate measures.
3. In serious cases of fraud, the Board of Examiners can propose to UM’s Executive Board that the student(s) concerned be permanently deregistered from the programme.
4. The Rules and Regulations include further provisions about what constitutes fraud and which disciplinary measures the Board of Examiners can impose.

Article 5.16a Invalid exam
If an exam involves irregularities that make it impossible to accurately assess the candidate’s knowledge, insight and skills, the Board of Examiners may declare the exam invalid for both the examinee and a group of examinees.

Article 5.17 Unsuitability (Iudicium Abeundi)
1. In exceptional cases and after careful consideration of the interests involved, the board of examination or the dean/the faculty board may ask the executive board to terminate or, as the case may be, refuse the enrolment of a student in a programme, if that student, through his behaviour or opinions ventured, has demonstrated his unsuitability for the practice of one or more professions for which he is trained by the programme he follows, or, as the case may be, for the practical preparation for the practice of the profession.
2. The relevant clauses of Maastricht University’s Enrolment Provisions apply.

Section 6 | Examination

Article 6.1 Examination
1. The Board of Examiners determines the result and date of the examination and issues the certificate as referred to in Article 6.3 as soon as the student has satisfied the requirements for the examination programme.
2. Prior to determining the result of the examination, the Board of Examiners may conduct their own investigation of the student’s knowledge regarding one or more components or aspects of the programme if and insofar as the results of the relevant exams give reason to do this.
3. To pass the examination, the student must pass all components referred to in Article 3.12.
4. To pass the examination and receive the certificate, the student must also have been registered for the programme during the period that the tests were taken.
5. A certificate may only be issued after it has been shown that the student has satisfied all the obligations, including paying the tuition fees.
6. The last day of the month in which the student satisfied all the examination obligations will be considered the examination date (graduation date).
7. Students who have passed the examination and who are entitled to the issuance of a certificate may, stating reasons, ask the Board of Examiners not to do this yet. This request must be submitted at least one month before the final assignment is turned in or the final test is taken. The Board of Examiners in any event grants the request - if the student is selected by the faculty for a double degree, an extracurricular internship or an extracurricular exchange, or - if the student holds or has held/will hold a board position for which of at least nine months of financial support is awarded from the profilingfund or holds or will hold a Student Introduction Committee (an 'INKOM') board position. The Board of Examiners may grant the request if refusal would result in an exceptional case of extreme unfairness because of the fact that the student concerned could not have taken the automatic graduation into account when he/she was planning his/her study.

Article 6.2 Degree
Students who have passed the examination will be awarded the degree 'Master of Science in Artificial Intelligence' or 'Master of Science in Data Science for Decision Making'.

Article 6.3 Certificate and statements
1. As proof that the examination was passed, the Board of Examiners issues a certificate, after it has been stated by or on behalf of UM’s Executive Board that the procedural requirements for receiving the certificate have been met. The certificate is based on the model that UM’s Executive Board has adopted. One certificate will be issued per programme, even if the student completes several programmes.
2. The certificate that the examination has been passed also indicates:
   a. the name of the institution;
   b. the name of the programme;
   c. the examination components;
   d. (if applicable) the right to practice a specific profession which is related to the certificate;
   e. the degree awarded;
   f. the date on which the programme was most recently accredited or was subjected to the new programme test.;
3. Students who are entitled to the issuance of a certificate may, stating reasons, ask the Board of Examiners not to do this yet (pursuant to Article 6.1(7)).
4. The certificate is signed by the chair of the Board of Examiners and the dean of the faculty.
5. The certificate includes a list of the examination components.
6. The board of examiners includes a diploma supplement as referred to in Article 7.11(4) of the Act with the certificate. This diploma supplement is based on the model adopted by UM’s Executive Board, which is in compliance with the agreed European standard format.
7. The Board of Examiners may award the ‘cum laude’ or ‘summa cum laude’ designation in accordance with the provisions in the Rules and Regulations.
8. Students who have passed more than one exam and who cannot be issued a certificate will upon request, receive a statement issued by the Board of Examiners which at least indicates the exams which they passed.

Article 6.4 Grade point average (GPA)
The diploma supplement referred to in Article 6.3(7) indicates the final grade point
average (GPA) as specified in the Rules and Regulations, to provide a reflection of the student’s academic performance.

Article 6.5 Right of appeal
Within six weeks after the decision by the examiner and the Board of Examiners is announced, the student may appeal this decision to UM’s Complaint Service Point. The appeal must be signed, must include a date and the name and address of the party lodging the appeal, must indicate the grounds for the appeal and, if possible, must include a copy of the decision being appealed.

Section 7 | Study Guidance

Article 7.1 Study progress administration
The faculty records the students’ individual study results and makes them available through the Student Portal.

Article 7.2 Study guidance
1. The faculty will provide the introduction and study guidance for students registered for the programme.
2. The study guidance includes
   i. an introduction during the first week of the first semester of the first academic year;
   ii. group and individual advice on possible study paths in and outside the programme, partly with a view to the professional options after the master’s programme.

Section 8 | Transitional and Final Provisions

Article 8.1 Amendments
1. Amendments to these regulations may be adopted in a separate decision by the faculty board, after a recommendation and consent from the programme committee and after consent from or consultation with the faculty council.
2. An amendment in these regulations will not pertain to the current academic year, unless the interests of the students will not reasonably be harmed as a result.
3. In addition, amendments may not affect, to the students’ detriment, a decision regarding a student which has been taken by the board of examiners pursuant to these regulations.

Article 8.2 Notice
1. The faculty board ensures that proper notice is given of these regulations, the rules and regulations adopted by the Board of Examiners, and any changes to these documents, by, for example, placing such notice on the faculty website/ and/or the Student Portal.
2. Any interested party may obtain a copy of the documents referred to in the first paragraph from the Student Affairs Office.

Article 8.2a Evaluation
The Faculty Board will ensure that the education of the programme is regularly evaluated, assessing at least – for the purpose of monitoring and if necessary
adapting the student workload – the amount of time students need to complete their duties as set out therein.

**Article 8.3 Unforeseen cases/safety net scheme**
1. In cases not covered or not clearly covered by these regulations, decisions are taken by or on behalf of the faculty board, after it has consulted with the board of examiners.
2. In individual cases in which application of the Education and Examination Regulations, would lead to manifestly unreasonable results, the board of examiners can deviate from the stated regulations in the student’s favour.

**Article 8.4 Effective date**
This Regulation will come into force on 1 September 2020 and will apply for the academic year 2020/2021.

Adopted by the faculty board on May 19, 2020.
## Appendix 1a: Master programme Artificial Intelligence

### Year 1

<table>
<thead>
<tr>
<th>Block</th>
<th>Course title</th>
<th>ECTS</th>
</tr>
</thead>
<tbody>
<tr>
<td>Block 1</td>
<td>Foundations of Agents</td>
<td>6</td>
</tr>
<tr>
<td></td>
<td>Intelligent Search &amp; Games</td>
<td>6</td>
</tr>
<tr>
<td>Block 2</td>
<td>Multi-Agent Systems</td>
<td>6</td>
</tr>
<tr>
<td></td>
<td>Advanced Concepts in Machine Learning</td>
<td>6</td>
</tr>
<tr>
<td>Block 3</td>
<td>Research project</td>
<td>6</td>
</tr>
<tr>
<td>Block 4</td>
<td>Autonomous Robotic Systems</td>
<td>6</td>
</tr>
<tr>
<td></td>
<td>1 elective course from the following set:</td>
<td></td>
</tr>
<tr>
<td></td>
<td>• Algorithms for Big Data</td>
<td>6</td>
</tr>
<tr>
<td></td>
<td>• Dynamic Game Theory</td>
<td></td>
</tr>
<tr>
<td></td>
<td>• Computational Statistics</td>
<td></td>
</tr>
<tr>
<td></td>
<td>• Advanced Natural Language Processing</td>
<td></td>
</tr>
<tr>
<td>Block 5</td>
<td>2 elective courses from the following set:</td>
<td></td>
</tr>
<tr>
<td></td>
<td>• Information Retrieval and Text Mining</td>
<td>6 + 6</td>
</tr>
<tr>
<td></td>
<td>• Deep Learning**</td>
<td></td>
</tr>
<tr>
<td></td>
<td>• Planning and Scheduling</td>
<td></td>
</tr>
<tr>
<td></td>
<td>• Computer Vision</td>
<td></td>
</tr>
<tr>
<td>Block 6</td>
<td>Research Project</td>
<td>6</td>
</tr>
</tbody>
</table>

### Year 2

#### Semester 1 Electives

- Internship
- Study Abroad
- Elective courses at other UM MSc programmes (at most 13 ECTS):
- Elective courses at DKE (for those started in September)*:
  - Block 1: 2 elective courses from the following set
    - Data Mining
    - Signal and Image Processing
    - Mathematical Optimization
    - Stochastic Decision Making
  - Block 2: 2 elective courses from the following set
    - Model Identification and Data Fitting
    - Applications of Image & Video Processing
    - Information Security
  - Block 3:
    - Research Project
- Elective courses (for those started in February)*:
  - Block 4: 2 elective courses from the following set
    - Algorithms for Big Data
    - Building & Mining Knowledge Graphs
    - Dynamic Game Theory
    - Computational Statistics
    - Advanced Natural Language Processing
  - Block 5: 2 elective courses from the following set
    - Information Retrieval and Text Mining
    - Symbolic Computation and Control
    - Planning and Scheduling
    - Computer Vision

<table>
<thead>
<tr>
<th>Block</th>
<th>Course title</th>
<th>ECTS</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Internship</td>
<td>30</td>
</tr>
</tbody>
</table>

*Elective courses are subject to availability and may vary from year to year.*
- Algorithms for Data Visualization
- Deep Learning**

**Passed components can only count for one particular year**

**To register for Deep Learning, the course Advanced Concepts in Machine Learning has to be passed.**

### Semester 2

<table>
<thead>
<tr>
<th>Master’s Thesis</th>
<th>ECTS</th>
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</thead>
<tbody>
<tr>
<td></td>
<td>30</td>
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### Appendix 1b: Master programme Data Science for Decision Making

#### Year 1

<table>
<thead>
<tr>
<th>Block 1</th>
<th>Data Mining</th>
<th>ECTS</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>1 elective course from the following set</td>
<td>6</td>
</tr>
<tr>
<td></td>
<td>Signal and Image Processing</td>
<td>6</td>
</tr>
<tr>
<td></td>
<td>Mathematical Optimization</td>
<td></td>
</tr>
<tr>
<td></td>
<td>Stochastic Decision Making</td>
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</table>

<table>
<thead>
<tr>
<th>Block 2</th>
<th>Model Identification and Data Fitting</th>
<th>ECTS</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>1 elective course from the following set</td>
<td>6</td>
</tr>
<tr>
<td></td>
<td>Advanced Concepts in Machine Learning</td>
<td></td>
</tr>
<tr>
<td></td>
<td>Applications of Image &amp; Video Processing</td>
<td></td>
</tr>
<tr>
<td></td>
<td>Information Security</td>
<td></td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Block 3</th>
<th>Research project</th>
<th>ECTS</th>
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<tbody>
<tr>
<td></td>
<td></td>
<td>6</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Block 4</th>
<th>Algorithms for Big Data</th>
<th>ECTS</th>
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</thead>
<tbody>
<tr>
<td></td>
<td>1 elective course from the following set</td>
<td>6</td>
</tr>
<tr>
<td></td>
<td>Building &amp; Mining Knowledge Graphs</td>
<td></td>
</tr>
<tr>
<td></td>
<td>Dynamic Game Theory</td>
<td></td>
</tr>
<tr>
<td></td>
<td>Computational Statistics</td>
<td></td>
</tr>
<tr>
<td></td>
<td>Advanced Natural Language Processing</td>
<td></td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Block 5</th>
<th>Planning and Scheduling</th>
<th>ECTS</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>1 elective course from the following set</td>
<td>6</td>
</tr>
<tr>
<td></td>
<td>Symbolic Computation and Control</td>
<td></td>
</tr>
<tr>
<td></td>
<td>Information Retrieval and Text Mining</td>
<td></td>
</tr>
<tr>
<td></td>
<td>Computer Vision</td>
<td></td>
</tr>
<tr>
<td></td>
<td>Algorithms for Data Visualization</td>
<td></td>
</tr>
<tr>
<td></td>
<td>Deep Learning**</td>
<td></td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Block 6</th>
<th>Research project</th>
<th>ECTS</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td>6</td>
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</tbody>
</table>

#### Year 2

**Semester 1 Electives**

- Internship
- Study Abroad
- Elective courses at other UM MSc programmes (at most 13 ECTS)
- Elective courses (for those started in September)*:
  - Block 1: 2 elective courses from the following set
• Signal and Image Processing
• Mathematical Optimization
• Stochastic Decision Making
• Foundation of Agents
• Intelligent Search and Games
- Block 2: 2 elective courses from the following set
• Multi Agent Systems
• Advanced Concepts in Machine Learning
• Applications of Image & Video Processing
• Information Security
- Block 3:
• Research Project

• Elective courses (for those started in February)*:
- Block 4: 2 elective courses from the following set
• Autonomous Robotic Systems
• Building & Mining Knowledge Graphs
• Dynamic Game Theory
• Computational Statistics
• Advanced Natural Language Processing
- Block 5: 2 elective courses from the following set
• Symbolic Computation and Control
• Information Retrieval and Text Mining
• Computer Vision
• Algorithms for Data Visualization
• Deep Learning**
- Block 6: Research Project

*Passed components only count for one particular year
**To register for Deep Learning, the course Advanced Concepts in Machine Learning has to be passed.

<table>
<thead>
<tr>
<th>Semester 2</th>
<th>ECTS</th>
</tr>
</thead>
<tbody>
<tr>
<td>Master’s Thesis</td>
<td>30</td>
</tr>
</tbody>
</table>

Appendix 2: Blocks with limited capacity (see Article 4.5)

<table>
<thead>
<tr>
<th>Block</th>
<th>Faculty</th>
<th>Number of participants</th>
<th>Offered in semester</th>
</tr>
</thead>
<tbody>
<tr>
<td>2.1-2.3</td>
<td>FSE/DKE</td>
<td>t.b.a. - Study Abroad</td>
<td>Semester 1 of year 2</td>
</tr>
</tbody>
</table>

Appendix 3: Cancelled

Appendix 4: Regulations for exchange programmes to study abroad

1. Components attained in connection with an exchange programme adopted by the faculty board may be incorporated into the examination for the programme for which the student is registered.
2. In principle, a student pursuing part of his/her studies in a foreign country does this
as part of the DKE’s exchange programme at one of its exchange partners.

3. Studies in connection with the exchange programme constitute part of the studies at UM.

4. The certificate may not be issued until the exchange programme has been completed and the results in the foreign country have been incorporated. The results will be taken into account for classification on the same basis as the results attained in the faculty.

5. In connection with the requirements regarding the level and content of the examination for the programme concerned, several conditions apply to incorporation. The Board of Examiners assesses whether a component may be incorporated into the examination. Incorporation is only possible with the Board of Examiners’ permission. The following conditions must be satisfied:
   a. the component is academic in nature;
   b. in terms of content, the component does not overlap with other examination components (Article 3.6 of the Education and Examination Regulations for Bachelor’s Programmes);
   c. the component was attained in connection with an exchange programme, except for the situation described below;
   d. the student requested written permission for incorporation beforehand, in principle, by submitting a request with appendices to the Student Affairs Office, which forwarded the information to the Board of Examiners;
   e. the student will submit additional information to the Board of Examiners upon request.

6. Point 2 may be deviated from if the student has good substantive reasons for studying in a foreign country at a university, which is not an exchange partner. Whether a foreign university is an exchange partner during the current year may be ascertained in the student handbook. The following procedure applies in such a situation:
   The student must timely submit (by letter or e-mail) a written request, stating reasons, to the Study Advisor. In addition to the student’s personal and study progress information, the request must include information about the university where the student wishes to study, as well as about the student’s substantive reasons for the request. These reasons must relate to the student’s own programme.
   The request is assessed by the Director of Studies. In this assessment, the level and the available courses of the university concerned are compared to the exchange partners’; the Director of Studies does not provide a judgment about the student’s specific choice of subject is not be provided, but only about general questions regarding the level and so forth.
   If the judgment about the university (regarding its level or otherwise) is favourable, the Study Advisor will notify the student and the Board of Examiners. The student will then follow the normal exchange programme procedure and will also follow the normal approval procedure with the Board of Examiners for incorporation of subjects.
   All cost involved for studying at a non-exchange partner is on behalf of the student.

7. The following rules apply to incorporation by the Board of Examiners of results attained in foreign countries:
   a) results must be provided to the Board of Examiners through clear, original written notices by the university concerned. In addition to personal information about the student and information about the university, these notices should
preferably include the following information: subject code and name; subject level; subject study load; and exam date and result; 
b) incorporation will only be possible if there has been a sufficient assessment. A mere statement that the subject was ‘taken’ or words to that effect will not be enough.
c) A result which constitutes a pass in the country concerned will also be considered a pass here, and a fail in a foreign country will also be considered a fail here; 
d) the study load conferred by the foreign university will be followed here without adjustment. If, for example, 5 ECTS are conferred by the university concerned for the subject, the subject will also count for 5 ECTS at the faculty, regardless of how difficult or easy the subject was for the student.
e) if the study load in the foreign country is not expressed in ECTS units but in other units, the Board of Examiners will convert the study load into ECTS units. The conversion will be based on the study load for a full year of study or a full programme in relation to a study load of 60 ECTS for a year at UM or 180 ECTS for a three-year bachelor’s programme.

8. Study delay caused by the study abroad is the responsibility of the student.
9. Insofar as the Education and Examination Regulations and this appendix do not address matters relating to incorporation of components attained elsewhere, the Board of Examiners will decide on the matter.

Appendix 5: Accountability for the language of instruction

The choice for the language of instruction of the programme is in line with the UM Code of Conduct on language in accordance with the Dutch Higher Education and Research Act (WHW) art. 7.2.

Because of the specific educational nature and profile of the programme, teaching and examinations are conducted in English. This guarantees the quality of education, because:

• The content of the programme has an international orientation and focus. Language of main course materials (handbooks, papers, tools) discussing or supporting the latest advances in AI or Data Science are in English.
• The academic community is internationally oriented and the staff is international. Project-Centred Learning implements the international classroom concept, which requires a common international language. Students participate in international collaboration by exchange programmes to study abroad, or conduct their internship and thesis at companies/institutes where English is the lingua franca.
• The labour market demand is internationally oriented (English speaking). Alumni typically end up in jobs at companies/institutes where English is the main language.
• The student intake and current population is internationally diverse and English is the common language (26% Dutch; 74% non-Dutch in 2017-2018).
5.2 Code of Conduct Department of Data Science and Knowledge Engineering

1. General Principles
With this Code of Conduct the Department of Data Science and Knowledge Engineering (DKE) aims to express its expectation of students’ behaviour in respect of personal and academic matters. This code of conduct applies to all students, regular and exchange students that take part in one of our programmes on the BA and MA level. The Department recognizes and values the diversity of student experiences and expectations, and is committed to treating students, both academically and personally, in a fair and transparent manner. We expect our students to comply with the requirements set down in this Code of Conduct. DKE expresses its commitment to:

• high academic standards, intellectual rigour and a high quality education;
• recognition of the importance of ideas and the pursuit of critical and open inquiry;
• tolerance, honesty and respect as the hallmarks of relationships throughout the University community; and
• high standards of ethical behaviour. All students are required to be aware of and act consistently with these values.

2. Personal conduct
All students are expected to:
• treat all employees, volunteers, peers or any other members of the public with politeness, honesty and respect;
• maintain a cooperative and collaborative approach to inter-personal relationships;
• respect the privacy of employees and other students
• be alert to undesirable or unlawful behaviour
• have respect for one another at all times. Behaviour such as hostility, violence, sexual harassment and discrimination on any grounds is not acceptable.
• be honest when taking part in academic (group) work. Taking exams, writing assignments and writing a thesis occurs independently and with academic integrity, refraining from fraud or improper citation
• show commitment by actively participating during lectures, classes and project work.

3. Communication
Communication among students or between students and staff members will be conducted in a polite and respectful manner. The Department informs the students via the official channels. These sources need to be checked daily. It cannot be argued by a student that he or she was unaware of information if it was disseminated via these official channels in a correct and timely manner. Furthermore:
• E-mail is a formal means of communication, which requires politeness and mutual respect.
• In case of a complaint and/or general or specific disappointment about an issue, students must remain polite in expressing those emotions.
• Students should refrain from sending personal information of staff members and peers to third parties.
• Problems among students or between students and staff members should be resolved in a civil manner and the parties involved are recommended to inform the student adviser.
4. Academic conduct related to courses and exams

All students are expected to:

• be in time for classes and exams. Students who are late for the exam will not be granted access.
• stay in the classroom/lecture hall during teaching. The obvious exception is a bathroom break. However, students are expected to show respect to the lecturer and their peers by not leaving and entering the room, which causes undesired disturbance.
• leave the classrooms and common rooms behind in an appropriate manner. Eating should occur outside the classroom. Eating and drinking is permitted in the commons rooms, as long as these rooms are maintained. This includes making sure that no waste is left behind.
• Refrain from disrespectful behaviour during class/lectures. This includes gaming and all other activities that are not part of the course programme.
• enrol timely and within the set deadlines for courses and exams (including resits). It is a student's responsibility to maintain current information in the student information system, and observe key dates and deadlines;
• read all official correspondence from the University, including email about the courses and exams
• act ethically and honestly in the preparation, conduct, submission and publication of academic work, and during all forms of assessment, including formal examinations and (in)formal tests
• behave professionally, ethically and respectfully in all dealings with partner Universities during the free semester and during Internships.
• During a semester abroad or an internship, DKE student represent the Department at their host organization and needs to be aware of both the rules and regulations at DKE about a semester abroad or an internship as well as the rules and regulations at the host institute. Students must always make sure to communicate politely and swiftly about any possible question or problem.
Staff

Dr. Stelios Asteriadis
Dr. Pieter Bonizzi
Dr. Alexia Brassouli
Dr. Cameron Browne

Dr. Rachel Cavill
Dr. Pieter Collins
Dr. ir. Kurt Driessens
Dr. Joël Karel

Dr. Steven Kelk
Dr. Siamak Mehrkanoon
Dr. Matúš Mihalák
Dr. Rico Möckel

Dr. Jan Niehues
Prof. dr. Ralf Peeters
Tom Pepels
Dr. Dimitris Rafailidis
Dr. ir. Nico Roos
Dr. Gijs Schoenmakers
Prof. dr. ir. Jan Scholtes
Dr. Katharina Schüller

Dr. Christof Seiler
Dr. Evgueni Smirnov
Dr. Jerry Spanakis
Dr. Katerina Stankova

Dr. Georgios Stamoulis
Prof. dr. Frank Thuijsman
Dr. Jos Uiterwijk
Prof. dr. Gerhard Weiss

Prof. Dr. Mark Winands
Dr. Steven Chaplick
Dr. Marieke Musegaas
Staff - continued

Dr. Enrique Hortal
Dr. Otti D’Huys
Dr. Chiara Sironi
Prof. Dr. Nava Tintarev
Prof. Dr. Anna Wilbik
Academic year 2020-2021

Bachelor’s Programme Data Science and Artificial Intelligence (year 1, 2, 3) and
Master’s Programmes Artificial Intelligence and Data Science for Decision Making: (year 1, 2)

Inkom Maastricht University
August 2020

Education periods
Period 1: 31 August - 16 October 2020
Period 2: 26 October - 11 December 2020
Period 3: 4 January - 22 January 2021
Period 4: 1 February - 26 March 2021
Period 5: 5 April - 28 May 2021
Period 6: 7 June - 25 June 2021

Introduction day
24, 25 August 2020 (Bachelor) and
27 August 2020 (Masters)
1 Feb 2021 (Masters, Feb intake)

Exam and Resit periods
Period 1 - Exams: 19 - 23 October 2020
Period 2 - Exams: 14 - 18 December 2020
Period 3 - Resits semester 1: 25 - 29 January 2021
Period 4 - Exams: 29 March - 2 April 2021
Period 5 - Exams: 31 May - 4 June 2021
Resits bachelor Year 3, Period 4: 31 May - 4 June 2021
Period 6 - Resits semester 2: 28 June - 2 July 2021

Project weeks
Period 3: Project weeks 4 - 22 January 2021
BA year 1, Final presentation: January 2021
BA year 2, Final presentation: January 2021
BA year 3, Final presentation: January 2021
MA year 1, Project seminar: January 2021
Period 6: Project weeks 7 - 25 June 2021
BA year 1, Final presentation: June 2021
BA year 2, Final presentation: June 2021
MA year 1, Project seminar: June 2021
Bachelor Winter Conference: December 2020
Bachelor Summer Conference: June 2021

Graduation
t.b.d. July 2021 (Masters)
t.b.d. July 2021 (Bachelor)

DKE students will have to register for their own courses, exams and resits through the
Student Portal.
Note: exams can also be scheduled in the evening, from 06.00 p.m. until 09.00 p.m.
DKE’s bachelor’s programme in Data Science and Artificial Intelligence has been ranked no.1 and is awarded the seal of Top Quality Programme three consecutive years in a row by Keuzegids Hoger Onderwijs.