The curriculum of the master programme is divided into four sub-curricula, each corresponding to one of the four main areas of competence in research of the Bonn Institute of Computer Science:

1. Algorithmics
2. Graphics, Vision, Audio
3. Information and Communication Management
4. Intelligent Systems

Module numbers MA-INF ASXY have been assigned according to the following key: vergeben:

- A = number of the area of competence
- S = semester within the master curriculum
- XY = sequential number within the semester and the respective area of competence (two digits)

According to the curriculum, all modules ought to be taken between the first and the third semester. The fourth semester is reserved for preparing the master thesis.

Contents

1 Algorithmics  
2 Graphics, Vision, Audio  
3 Information and Communication Management  
4 Intelligent Systems  
5 Master Thesis
## 1 Algorithmics

<table>
<thead>
<tr>
<th>Code</th>
<th>Type</th>
<th>CP</th>
<th>Title</th>
</tr>
</thead>
<tbody>
<tr>
<td>MA-INF 1102</td>
<td>L4E2</td>
<td>9</td>
<td>Combinatorial Optimization</td>
</tr>
<tr>
<td>MA-INF 1103</td>
<td>L4E2</td>
<td>9</td>
<td>Cryptography</td>
</tr>
<tr>
<td>MA-INF 1105</td>
<td>L2E2</td>
<td>6</td>
<td>Algorithms for Data Analysis</td>
</tr>
<tr>
<td>MA-INF 1107</td>
<td>L2E2</td>
<td>6</td>
<td>Foundations of Quantum Computing</td>
</tr>
<tr>
<td>MA-INF 1108</td>
<td>L2E2</td>
<td>6</td>
<td>Introduction to High Performance Computing: Architecture Features and Practical Parallel Programming</td>
</tr>
<tr>
<td>MA-INF 1201</td>
<td>L4E2</td>
<td>9</td>
<td>Approximation Algorithms</td>
</tr>
<tr>
<td>MA-INF 1202</td>
<td>L4E2</td>
<td>9</td>
<td>Chip Design</td>
</tr>
<tr>
<td>MA-INF 1203</td>
<td>L4E2</td>
<td>9</td>
<td>Discrete and Computational Geometry</td>
</tr>
<tr>
<td>MA-INF 1205</td>
<td>6</td>
<td></td>
<td>Graduate Seminar Discrete Optimization</td>
</tr>
<tr>
<td>MA-INF 1206</td>
<td>Sem2</td>
<td>4</td>
<td>Seminar Randomized and Approximation Algorithms</td>
</tr>
<tr>
<td>MA-INF 1207</td>
<td>Lab4</td>
<td>9</td>
<td>Lab Combinatorial Algorithms</td>
</tr>
<tr>
<td>MA-INF 1209</td>
<td>Sem2</td>
<td>4</td>
<td>Seminar Advanced Topics in Cryptography</td>
</tr>
<tr>
<td>MA-INF 1213</td>
<td>L4E2</td>
<td>9</td>
<td>Randomized Algorithms and Probabilistic Analysis</td>
</tr>
<tr>
<td>MA-INF 1217</td>
<td>Sem2</td>
<td>4</td>
<td>Seminar Theoretical Foundations of Data Science</td>
</tr>
<tr>
<td>MA-INF 1218</td>
<td>L4E2</td>
<td>9</td>
<td>Algorithms and Uncertainty</td>
</tr>
<tr>
<td>MA-INF 1219</td>
<td>Sem2</td>
<td>4</td>
<td>Seminar Algorithmic Game Theory</td>
</tr>
<tr>
<td>MA-INF 1220</td>
<td>Sem2</td>
<td>4</td>
<td>Seminar Algorithms for Computational Analytics</td>
</tr>
<tr>
<td>MA-INF 1221</td>
<td>Lab4</td>
<td>9</td>
<td>Lab Computational Analytics</td>
</tr>
<tr>
<td>MA-INF 1222</td>
<td>Lab4</td>
<td>9</td>
<td>Lab High Performance Optimization</td>
</tr>
<tr>
<td>MA-INF 1223</td>
<td>L4E2</td>
<td>9</td>
<td>Privacy Enhancing Technologies</td>
</tr>
<tr>
<td>MA-INF 1224</td>
<td>L2E2</td>
<td>5</td>
<td>Quantum Computing Algorithms</td>
</tr>
<tr>
<td>MA-INF 1225</td>
<td>Lab4</td>
<td>9</td>
<td>Lab Exploring HPC technologies</td>
</tr>
<tr>
<td>MA-INF 1301</td>
<td>L4E2</td>
<td>9</td>
<td>Algorithmic Game Theory</td>
</tr>
<tr>
<td>MA-INF 1304</td>
<td>Sem2</td>
<td>4</td>
<td>Seminar Computational Geometry</td>
</tr>
<tr>
<td>MA-INF 1305</td>
<td>6</td>
<td></td>
<td>Graduate Seminar on Applied Combinatorial Optimization</td>
</tr>
<tr>
<td>MA-INF 1307</td>
<td>Sem2</td>
<td>4</td>
<td>Seminar Advanced Algorithms</td>
</tr>
<tr>
<td>MA-INF 1308</td>
<td>Lab4</td>
<td>9</td>
<td>Lab Algorithms for Chip Design</td>
</tr>
<tr>
<td>MA-INF 1309</td>
<td>Lab4</td>
<td>9</td>
<td>Lab Efficient Algorithms: Design, Analysis and Implementation</td>
</tr>
<tr>
<td>MA-INF 1314</td>
<td>L4E2</td>
<td>9</td>
<td>Online Motion Planning</td>
</tr>
<tr>
<td>MA-INF 1315</td>
<td>Lab4</td>
<td>9</td>
<td>Lab Computational Geometry</td>
</tr>
<tr>
<td>MA-INF 1316</td>
<td>Lab4</td>
<td>9</td>
<td>Lab Cryptography</td>
</tr>
<tr>
<td>MA-INF 1320</td>
<td>Lab4</td>
<td>9</td>
<td>Lab Advanced Algorithms</td>
</tr>
<tr>
<td>MA-INF 1321</td>
<td>L2E2</td>
<td>6</td>
<td>Binary Linear and Quadratic Optimization</td>
</tr>
<tr>
<td>MA-INF 1322</td>
<td>Sem2</td>
<td>4</td>
<td>Seminar Focus Topics in High Performance Computing</td>
</tr>
<tr>
<td>MA-INF 1323</td>
<td>L4E2</td>
<td>9</td>
<td>Computational Topology</td>
</tr>
</tbody>
</table>
MA-INF 1102 Combinatorial Optimization

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>at least every year</td>
</tr>
</tbody>
</table>

**Module coordinator**
Prof. Dr. Jens Vygen

**Lecturer(s)**
All lecturers of Discrete Mathematics

**Programme**
M. Sc. Computer Science

**Mode**
Optional

**Semester**
1. or 2.

**Learning goals: technical skills**
Advanced knowledge of combinatorial optimization. Modelling and development of solution strategies for combinatorial optimization problems

**Learning goals: soft skills**
Mathematical modelling of practical problems, abstract thinking, presentation of solutions to exercises

**Contents**
Matchings, b-matchings and T-joins, optimization over matroids, submodular function minimization, travelling salesman problem, polyhedral combinatorics, NP-hard problems

**Prerequisites**
none

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td></td>
<td>4</td>
<td>60 T / 105 S</td>
<td>5.5</td>
</tr>
<tr>
<td>Exercises</td>
<td></td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

**Graded exams**
Oral exam

**Ungraded coursework (required for admission to the exam)**
Successful exercise participation

**Literature**
MA-INF 1103  Cryptography

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

Module coordinator | Lecturer(s)
Dr. Michael Nüsken | Dr. Michael Nüsken

Programme | Mode | Semester
M. Sc. Computer Science | Optional | 1. or 2.

Learning goals: technical skills
Understanding of security concerns and measures, and of the interplay between computing power and security requirements. Mastery of the basic techniques for cryptosystems and cryptanalysis

Learning goals: soft skills
Oral presentation (in tutorial groups), written presentation (of exercise solutions), team collaboration in solving homework problems, critical assessment

Contents
Basic private-key and public-key cryptosystems: AES, RSA, group-based. Security reductions. Key exchange, cryptographic hash functions, signatures, identification; factoring integers and discrete logarithms; lower bounds in structured models.

Prerequisites
none

Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td></td>
<td>4</td>
<td>60 T / 105 S</td>
<td>5.5</td>
</tr>
<tr>
<td>Exercises</td>
<td></td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
</tr>
</tbody>
</table>

Graded exams
Written exam

Ungraded coursework (required for admission to the exam)
The completion of regularly provided exercise sheets. Each student must present twice in the tutorial.

Literature
- Course notes
## MA-INF 1105  Algorithms for Data Analysis

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>180 h</td>
<td>6 CP</td>
<td>1 semester</td>
<td>at least every 2 years</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Module coordinator</th>
<th>Lecturer(s)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Prof. Dr. Petra Mutzel</td>
<td>Prof. Dr. Petra Mutzel</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Programme</th>
<th>Mode</th>
<th>Semester</th>
</tr>
</thead>
<tbody>
<tr>
<td>M. Sc. Computer Science</td>
<td>Optional</td>
<td>1. or 2.</td>
</tr>
</tbody>
</table>

### Learning goals: technical skills

Deeper insights into selected methods and techniques of modern algorithmics with respect to big data and/or analytics tasks.

### Learning goals: soft skills

Presentation of solutions and methods, critical discussion of applied methods and techniques.

### Contents

Advanced algorithmic techniques and data structures relevant to analytic tasks for big data, i.e., algorithms for graph similarity, parallel algorithms, I/O-data structures, and streaming algorithms.

### Prerequisites

**Required:**

none

**Recommended:**

Introductory knowledge of foundations of algorithms and data structures is essential.

### Course meetings

| Teaching format | Group size | h/week | Workload|h | CP | T = face-to-face teaching |
|----------------|------------|--------|----------|-----|--------------------------|
| Lecture        | 2          | 2      | 30 T / 45 S | 2.5 | S = independent study    |
| Exercises      | 2          | 2      | 30 T / 75 S | 3.5 |

### Graded exams

Oral exam

**Ungraded coursework (required for admission to the exam)**

Successful exercise participation
# MA-INF 1107 Foundations of Quantum Computing

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>180 h</td>
<td>6 CP</td>
<td>1 semester</td>
<td>every 2 years</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Module coordinator</th>
<th>Lecturer(s)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Prof. Dr.-Ing. Christian Bauckhage</td>
<td>Prof. Dr.-Ing. Christian Bauckhage</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Programme</th>
<th>Mode</th>
<th>Semester</th>
</tr>
</thead>
<tbody>
<tr>
<td>M. Sc. Computer Science</td>
<td>Optional</td>
<td>1. or 3.</td>
</tr>
</tbody>
</table>

## Learning goals: technical skills

Upon successful completion of this module, students should be able to describe fundamental concepts and techniques (qubits, quantum registers, quantum gates, quantum circuits) in quantum computing. Students will be equipped with specific, quantum computing related programming know-how; based on knowledge and skills acquired, students should be able to

- devise quantum computing algorithms for basic computational tasks
- run these algorithms on (simulated) quantum computers

## Learning goals: soft skills

In the exercises, students will have the opportunity to put their knowledge into practice, since they will realize small projects on computing with quantum gates and their solutions using quantum inspired methods or genuine quantum methods. This requires teamwork; upon successful completion of the module, students should be able to

- draft and implement basic quantum computing algorithms
- apply quantum computing (simulations) to test these algorithms
- prepare and give oral presentations about their work in front of an audience

## Contents

Boolean algebras and Boolean lattices; cellular automata; classical digital computing; classical reversible computing; mathematical foundations of quantum computing (complex vector spaces, tensor products, unitary operators, Hermitian operators, qubits, superposition, entanglement); quantum gate computing; quantum circuits

## Prerequisites

**Recommended:**
Good working knowledge of theory and practice of linear algebra

## Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td>2</td>
<td>30 T / 45 S</td>
<td>2.5</td>
<td></td>
</tr>
<tr>
<td>Exercises</td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
<td></td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

## Graded exams

Written exam

**Ungraded coursework (required for admission to the exam)**
Successful exercise participation

## Forms of media

- lecture slides / lecture notes are made available online
- notebooks with programming examples are made available online

## Literature

Introduction to High Performance Computing: Architecture Features and Practical Parallel Programming

Workload: 180 h
Credit points: 6 CP
Duration: 1 semester
Frequency: every year

Module coordinator: Prof. Dr. Estela Suarez
Lecturer(s): Prof. Dr. Estela Suarez

Programme: M. Sc. Computer Science
Mode: Optional

Learning goals: technical skills
Understanding principles of computer architecture of modern HPC systems at component (processor, accelerators) and system level (system architecture, network, memory hierarchy) and their implication for application programming. Ability to program parallel computers, employing multi-core and multi-node features. Programming CPU and GPUs. Understanding the quality of performance and scaling behaviour, and applying the measures needed to improve them.

Learning goals: soft skills
Ability to select a specific HPC topic and present it in a clear and comprehensive manner suitable for a lightning talk (10 min).

Contents
Computer architectures, system components (CPU, memory, network) and their interrelation.
Software environment
Access to HPC compute resources at the Jülich Supercomputing Centre
Practical use of parallel programming paradigms (MPI, OpenMP, CUDA)
Performance of applications and scaling behavior, understanding and strategies for improvement
Current challenges in HPC

Prerequisites
Required:
Knowledge of a modern programming language (ideally C/C++ and Python).
Interest in High Performance Computing
Cannot be taken after completing MA-INF 1106.

Recommended:
Bachelor lecture on computer architecture

Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td>2</td>
<td></td>
<td>30 T / 45 S</td>
<td>2.5</td>
</tr>
<tr>
<td>Exercises</td>
<td>2</td>
<td></td>
<td>30 T / 75 S</td>
<td>3.5</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

Graded exams
Written exam

Ungraded coursework (required for admission to the exam)
Successful participation in the exercises

Forms of media
Laptop and projector
Literature

- OpenMP Application Programming Interface, Version 4.5, November 2015
MA-INF 1201  Approximation Algorithms

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>at least every year</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Module coordinator</th>
<th>Lecturer(s)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Prof. Dr. Jens Vygen</td>
<td>All lecturers of Discrete Mathematics, Senior Prof. Dr. Marek Karpinski</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Programme</th>
<th>Mode</th>
<th>Semester</th>
</tr>
</thead>
<tbody>
<tr>
<td>M. Sc. Computer Science</td>
<td>Optional</td>
<td>2. or 3.</td>
</tr>
</tbody>
</table>

**Learning goals: technical skills**

Introduction to design and analysis of most important approximation algorithms for NP-hard combinatorial optimization problems, and various techniques for proving lower and upper bounds, probabilistic methods and applications

**Learning goals: soft skills**

Presentation of solutions and methods, critical discussion of applied methods and techniques

**Contents**


**Prerequisites**

**Recommended:**

Introductory knowledge of foundations of algorithms and complexity theory is essential.

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload</th>
<th>h</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td></td>
<td>4</td>
<td>60 T / 105 S</td>
<td>5.5</td>
<td></td>
</tr>
<tr>
<td>Exercises</td>
<td></td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
<td></td>
</tr>
</tbody>
</table>

**Graded exams**

<table>
<thead>
<tr>
<th>T = face-to-face teaching</th>
<th>S = independent study</th>
</tr>
</thead>
</table>

**Ungraded coursework (required for admission to the exam)**

The completion of regularly provided exercise sheets. The work can be done in groups of up to two students. A total of 50% of the points must be achieved. Each student must present a solution to an exercise in the exercise sessions twice.

**Literature**

MA-INF 1202  Chip Design

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

**Module coordinator**

Prof. Dr. Jens Vygen

**Lecturer(s)**

All lecturers of Discrete Mathematics

**Programme**

M. Sc. Computer Science

<table>
<thead>
<tr>
<th>Mode</th>
<th>Semester</th>
</tr>
</thead>
<tbody>
<tr>
<td>Optional</td>
<td>1. or 2.</td>
</tr>
</tbody>
</table>

**Learning goals: technical skills**

Knowledge of the central problems and algorithms in chip design. Competence to develop and apply algorithms for solving real-world problems, also with respect to technical constraints. Techniques to develop and implement efficient algorithms for very large instances.

**Learning goals: soft skills**

Mathematical modelling of problems occurring in chip design, development of efficient algorithms, abstract thinking, presentation of solutions to exercises

**Contents**

Problem formulation and design flow for chip design, logic synthesis, placement, routing, timing analysis and optimization

**Prerequisites**

none

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td>4</td>
<td>4</td>
<td>60 T / 105 S</td>
<td>5.5</td>
</tr>
<tr>
<td>Exercises</td>
<td>2</td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
</tr>
</tbody>
</table>

T = face-to-face teaching  
S = independent study

**Graded exams**

Oral exam

**Ungraded coursework (required for admission to the exam)**

The completion of regularly provided exercise sheets. The work can be done in groups of up to two students. A total of 50% of the points must be achieved. Each student must present a solution to an exercise in the exercise sessions twice.

**Literature**

- S. Held, J. Vygen: Chip Design. Lecture Notes (distributed during the course)
MA-INF 1203  Discrete and Computational Geometry

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

Module coordinator
Prof. Dr. Anne Driemel
Lecturer(s)
Prof. Dr. Anne Driemel, PD Dr. Elmar Langetepe, Dr. Herman Haverkort

Programme
M. Sc. Computer Science

Mode
Optional

Semester
1-4.

Learning goals: technical skills
Knowledge of fundamental theorems and concepts in the area of discrete and computational geometry; design and analysis of geometric algorithms; combinatorial analysis of the complexity of geometric configurations; to apply this knowledge autonomously in solving new problems.

Learning goals: soft skills
Social competence (communication, presenting one’s own solutions, goal-oriented discussions in teams), methodical competence (analysis, abstraction, proofs), individual competence (commitment and willingness to learn, creativity, endurance).

Contents
Fundamentals of convex sets, Voronoi diagrams, hyperplane arrangements, well-separated pair decomposition, spanners, metric space embedding, dimension reduction, VC-dimension, epsilon-nets, visibility, point location, range searching, randomized incremental construction, geometric distance problems in dimension two and higher.

Prerequisites
Recommended:
BA-INF 114 – Grundlagen der algorithmischen Geometrie

Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td>4</td>
<td>4</td>
<td>60 T / 105 S</td>
<td>5.5</td>
</tr>
<tr>
<td>Exercises</td>
<td>2</td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
</tr>
</tbody>
</table>

Graded exams
Oral exam

Ungraded coursework (required for admission to the exam)
Successful exercise participation

Literature
- Narasimhan/Smid, Geometric Spanner Networks
- Klein, Concrete and Abstract Voronoi Diagrams
MA-INF 1205  Graduate Seminar Discrete Optimization

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>180 h</td>
<td>6 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

**Module coordinator**
Prof. Dr. Jens Vygen

**Lecturer(s)**
All lecturers of Discrete Mathematics

**Programme**
M. Sc. Computer Science

**Mode**
Optional

**Semester**
2.

**Learning goals: technical skills**
Competence to understand new research results based on original literature, to put such results in a broader context and present such results and relations.

**Learning goals: soft skills**
Ability to read and understand research papers, abstract thinking, presentation of mathematical results in a talk

**Contents**
A current research topic in discrete optimization will be chosen each semester and discussed based on original literature.

**Prerequisites**
Recommended:
MA-INF 1102 – Combinatorial Optimization

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Seminar</td>
<td>10</td>
<td>4</td>
<td>60 T / 120 S</td>
<td>6</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

**Graded exams**
Oral presentation, written report

**Ungraded coursework (required for admission to the exam)**

**Literature**
The topics and the relevant literature will be announced towards the end of the previous semester.
MA-INF 1206 Seminar Randomized and Approximation Algorithms

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>120 h</td>
<td>4 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

**Module coordinator**

Prof. Dr. Heiko Röglin

**Lecturer(s)**

Prof. Dr. Anne Driemel, Prof. Dr. Thomas Kesselheim,
Prof. Dr. Heiko Röglin, PD Dr. Elmar Langetepe, Dr. Herman Haverkort,
Senior Prof. Dr. Marek Karpinski

**Programme**

M. Sc. Computer Science

**Mode**

Optional

**Semester**

2.

**Learning goals: technical skills**

Ability to perform individual literature search, critical reading, understanding, and clear presentation.

**Learning goals: soft skills**

Presentation of solutions and methods, critical discussion of applied methods and techniques

**Contents**

Current topics in design and analysis of randomized and approximation algorithms based on lastest research literature

**Prerequisites**

none

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Seminar</td>
<td>10</td>
<td>2</td>
<td>30 T / 90 S</td>
<td>4</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

**Graded exams**

Oral presentation, written report

**Ungraded coursework (required for admission to the exam)**

**Literature**

The relevant literature will be announced in time.
MA-INF 1207  Lab Combinatorial Algorithms

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

**Module coordinator**
Prof. Dr. Jens Vygen

**Lecturer(s)**
All lecturers of Discrete Mathematics

**Programme**
M. Sc. Computer Science

**Mode**
Optional

**Semester**
2.

**Learning goals: technical skills**
Competence to implement advanced combinatorial algorithms, handling nontrivial data structures, testing, documentation. Advanced software techniques.

**Learning goals: soft skills**
Efficient implementation of complex algorithms, abstract thinking, documentation of source code

**Contents**
Certain combinatorial algorithms will be chosen each semester. The precise task will be explained in a meeting in the previous semester.

**Prerequisites**
Recommended:
MA-INF 1102 – Combinatorial Optimization

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lab</td>
<td>8</td>
<td>4</td>
<td>60 T / 210 S</td>
<td>9</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

**Graded exams**
Oral presentation, written report

**Ungraded coursework (required for admission to the exam)**

**Literature**
The topics and the relevant literature will be announced towards the end of the previous semester.
### MA-INF 1209  Seminar Advanced Topics in Cryptography

<table>
<thead>
<tr>
<th><strong>Workload</strong></th>
<th><strong>Credit points</strong></th>
<th><strong>Duration</strong></th>
<th><strong>Frequency</strong></th>
</tr>
</thead>
<tbody>
<tr>
<td>120 h</td>
<td>4 CP</td>
<td>1 semester</td>
<td>every semester</td>
</tr>
</tbody>
</table>

**Module coordinator**

Dr. Michael Nüsken

**Lecturer(s)**

Dr. Michael Nüsken

**Programme**

M. Sc. Computer Science

**Mode**

Optional

**Semester**

2. or 3.

### Learning goals: technical skills

Understanding research publications, often written tersely. Distilling this into a presentation. Determination of relevant vs. irrelevant material. Developing a presentation that fascinates fellow students.

### Learning goals: soft skills

Understanding and presenting material both orally and in visual media. Motivating other students to participate. Critical assessment of research results.

### Contents

A special topic within cryptography, changing from year to year, is studied in depth, based on current research literature.

### Prerequisites

**Recommended:**

Basic knowledge in cryptography is highly recommended, eg. by MA-INF 1103 – Cryptography.

### Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Seminar</td>
<td>10</td>
<td>2</td>
<td>30 T / 90 S</td>
<td>4</td>
</tr>
</tbody>
</table>

T = face-to-face teaching  
S = independent study

### Graded exams

Oral presentation, written report

### Ungraded coursework (required for admission to the exam)

### Literature

Current cryptographic literature.
MA-INF 1213  Randomized Algorithms and Probabilistic Analysis

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

**Module coordinator**
Prof. Dr. Heiko Röglin

**Lecturer(s)**
Prof. Dr. Heiko Röglin

**Programme**
M. Sc. Computer Science

**Mode**
Optional

**Semester**
2. or 4.

**Learning goals: technical skills**
Understanding of models and techniques for the probabilistic analysis of algorithms as well as for the design and analysis of randomized algorithms

**Learning goals: soft skills**
Oral and written presentation of solutions and methods, abstract thinking

**Contents**
Design and analysis of randomized algorithms
- complexity classes
- Markov chains and random walks
- tail inequalities
- probabilistic method
- smoothed and average-case analysis
- simplex algorithm
- local search algorithms
- clustering algorithms
- combinatorial optimization problems
- multi-objective optimization

**Prerequisites**
none

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td>4</td>
<td>60 T / 105 S</td>
<td>5.5</td>
<td></td>
</tr>
<tr>
<td>Exercises</td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
<td></td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

**Graded exams**
Oral exam

**Ungraded coursework (required for admission to the exam)**
The completion of regularly provided exercise sheets. The work can be done in groups of up to three students. A total of 25% of the points must be achieved.

**Literature**
- lecture notes
- research articles
MA-INF 1217  Seminar Theoretical Foundations of Data Science

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>120 h</td>
<td>4 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

**Module coordinator**

Prof. Dr. Heiko Röglin

**Lecturer(s)**

Prof. Dr. Anne Driemel, Prof. Dr. Thomas Kesselheim, Prof. Dr. Heiko Röglin, PD Dr. Elmar Langetepe, Dr. Herman Haverkort

**Programme**

M. Sc. Computer Science

**Mode**

Optional

**Semester**

2. or 3.

**Learning goals: technical skills**

Ability to understand new research results presented in original scientific papers.

**Learning goals: soft skills**

Ability to present and to critically discuss these results in the framework of the corresponding area.

**Contents**

Current conference and journal papers

**Prerequisites**

none

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Seminar</td>
<td>10</td>
<td>2</td>
<td>30 T / 90 S</td>
<td>4</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

**Graded exams**

Oral presentation, written report

**Ungraded coursework (required for admission to the exam)**
MA-INF 1218  Algorithms and Uncertainty

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>at least every 2 years</td>
</tr>
</tbody>
</table>

Module coordinator
Prof. Dr. Thomas Kesselheim

Lecturer(s)
Prof. Dr. Thomas Kesselheim

Programme
M. Sc. Computer Science

Mode
Optional

Semester
2. or 3.

Learning goals: technical skills
Understanding approaches for modeling uncertainty in algorithmic theory. Designing and analyzing algorithms with performance guarantees in the context of uncertainty.

Learning goals: soft skills
Oral and written presentation of solutions and methods

Contents
- Advanced Online Algorithms
- Markov Decisions Processes
- Stochastic and Robust Optimization
- Online Learning Algorithms and Online Convex Optimization

Prerequisites
Recommended:
Solid background in algorithms, calculus, and probability theory. Specialized knowledge about certain algorithms is not necessary.

Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td>4</td>
<td>60 T / 105 S</td>
<td>5.5</td>
<td></td>
</tr>
<tr>
<td>Exercises</td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
<td></td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

Graded exams

Oral exam

Ungraded coursework (required for admission to the exam)
Successful exercise participation

Literature

lecture notes, research articles
## MA-INF 1219 Seminar Algorithmic Game Theory

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>120 h</td>
<td>4 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

### Module coordinator
Prof. Dr. Thomas Kesselheim

### Lecturer(s)
Prof. Dr. Thomas Kesselheim

### Programme
M. Sc. Computer Science

### Mode
Optional

### Semester
2. or 3.

### Learning goals: technical skills
Ability to understand new research results presented in original scientific papers.

### Learning goals: soft skills
Ability to perform individual literature search, critical reading, and clear didactic presentation

### Contents
Advanced topics in Algorithmic Game Theory and Algorithmic Mechanism Design based on current conference and journal papers

### Prerequisites
none

### Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Seminar</td>
<td>10</td>
<td>2</td>
<td>30 T / 90 S</td>
<td>4</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

### Graded exams
Oral presentation, written report

### Ungraded coursework (required for admission to the exam)
MA-INF 1220  Seminar Algorithms for Computational Analytics

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>120 h</td>
<td>4 CP</td>
<td>1 semester</td>
<td>at least every year</td>
</tr>
</tbody>
</table>

**Module coordinator**
Prof. Dr. Petra Mutzel

**Lecturer(s)**
Prof. Dr. Petra Mutzel

**Programme**
M. Sc. Computer Science

**Mode**
Optional

**Semester**
2. or 3.

**Learning goals: technical skills**
Ability to perform individual literature search, critical reading, understanding, and clear didactic presentation.

**Learning goals: soft skills**
Ability to present and to critically discuss these results in the framework of the corresponding area.

**Contents**
Current topics in algorithms for computational analytics based on recent research literature.

**Prerequisites**
Recommended:
Interest in Algorithms

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Seminar</td>
<td>10</td>
<td>2</td>
<td>30 T / 90 S</td>
<td>4</td>
</tr>
</tbody>
</table>

T = face-to-face teaching  
S = independent study

**Graded exams**
Oral presentation, written report

**Ungraded coursework (required for admission to the exam)**

**Literature**
The relevant literature will be announced in time.
**MA-INF 1221 Lab Computational Analytics**

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

**Module coordinator**
Prof. Dr. Petra Mutzel

**Lecturer(s)**
Prof. Dr. Petra Mutzel

**Programme**
M. Sc. Computer Science

**Mode**
Optional

**Semester**
2. or 3.

**Learning goals: technical skills**
Ability to design, analyze and implement efficient algorithms for computational analytics problems. The LAB also includes experimental evaluation and documentation of the implemented software.

**Learning goals: soft skills**
Ability to properly present and defend design decisions, to prepare readable documentation of software; skills in constructively collaborating with others in small teams over a longer period of time; ability to classify one’s own results into the state-of-the-art of the resp. area.

**Contents**
Design of efficient exact and approximate algorithms and data structures for computational analytics problems.

**Prerequisites**
Recommended: Interest in algorithms

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload</th>
<th>h</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lab</td>
<td>8</td>
<td>4</td>
<td>60 T / 210 S</td>
<td>9</td>
<td></td>
</tr>
</tbody>
</table>

T = face-to-face teaching  
S = independent study

<table>
<thead>
<tr>
<th>Graded exams</th>
</tr>
</thead>
<tbody>
<tr>
<td>Oral presentation, written report</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Ungraded coursework (required for admission to the exam)</th>
</tr>
</thead>
</table>

<table>
<thead>
<tr>
<th>Literature</th>
</tr>
</thead>
<tbody>
<tr>
<td>The relevant literature will be announced in time.</td>
</tr>
</tbody>
</table>
MA-INF 1222  Lab High Performance Optimization

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

**Module coordinator**
Prof. Dr. Petra Mutzel

**Lecturer(s)**
Prof. Dr. Petra Mutzel, Dr. Sven Mallach

**Programme**
M. Sc. Computer Science

**Mode**
Optional

**Semester**
2. or 3.

**Learning goals: technical skills**
Ability to design, analyze and implement algorithms for computational analytics and optimization problems. The lab also includes experimental evaluation and documentation of the implemented software.

**Learning goals: soft skills**
Ability to properly present and defend design decisions, to prepare readable documentation of software; skills in constructively collaborating with others in small teams over a longer period of time; ability to classify one’s own results into the state-of-the-art of the resp. area.

**Contents**

**Prerequisites**
none

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lab</td>
<td>8</td>
<td>4</td>
<td>60 T / 210 S</td>
<td>9</td>
</tr>
</tbody>
</table>

T = face-to-face teaching  
S = independent study

**Graded exams**
Oral presentation, written report

**Ungraded coursework (required for admission to the exam)**

**Literature**
The relevant literature will be announced in time.
MA-INF 1223  Privacy Enhancing Technologies

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

Module coordinator
Dr. Michael Nüsken

Lecturer(s)
Dr. Michael Nüsken

Programme
M. Sc. Computer Science

Mode
Optional

Semester
2. or 3.

Learning goals: technical skills
Knowledge: Cryptographic schemes for enhancing privacy, underlying security notions, applications and restrictions.
Skills: Secure application of sophisticated cryptographic schemes. Evaluation of their correctness, efficiency and security in an application setting.

Learning goals: soft skills

Contents
With more and more data available a clear separation of sensitive data is necessary and needs to be protected. Some of that data must stay within strict environments, for examples hospitals must store certain highly sensitive medical information about patients but they are not allowed to store it outside its own facilities. Some of that data is stored or collected in a cloud environment in encrypted form, say data from a medical device or a smart home. But it shall still be possible to derive important conclusions from it, for example to send immediate help to a patient suffering a heart attack.

Innovative solutions are needed in this area of tension. The research in cryptography provides some highly sophisticated tools for solving the like problems.

• Fully homomorphic encryption (FHE).
• Zero-Knowledge techniques, in particular: Non-interactive zero-knowledge proof (NIZKs).
• Secure multi-party computations (MPC).
• Anonymisation, TOR. Pseudonymization. Blinding.
• Weaker privacy notions, like differential privacy.

Prerequisites
Recommended:
Basic knowledge in cryptography is highly recommended.
A profound mathematical background does help. In particular, precise mathematical formulation and reasoning are important, but also topics like elementary number theory and discrete mathematics, especially lattices, are interesting.

Course meetings

<table>
<thead>
<tr>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td>4</td>
<td>60 T / 105 S</td>
<td>5.5</td>
</tr>
<tr>
<td>Exercises</td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

Graded exams
Written exam

Ungraded coursework (required for admission to the exam)
The completion of regularly provided exercise sheets. Each student must present twice in the tutorial.
MA-INF 1224  Quantum Computing Algorithms

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>150 h</td>
<td>5 CP</td>
<td>1 semester</td>
<td>every 2 years</td>
</tr>
</tbody>
</table>

Module coordinator: Prof. Dr. Christian Bauckhage  
Lecturer(s): Prof. Dr. Christian Bauckhage

Programme:  
M. Sc. Computer Science  
Mode: Optional  
Semester: 2. or 4.

**Learning goals: technical skills**

Upon successful completion of this module, students should be able to describe fundamental concepts behind working quantum algorithms.

Students acquire quantum computing programming know-how; based on knowledge and skills acquired, students should be able to:

- run quantum algorithms on (simulated) quantum computing platforms
- devise their own algorithms for optimization or classification problems that can be solved on quantum computers

**Learning goals: soft skills**

In the exercises, students can put their quantum computing knowledge into practice and realize small projects involving the implementation of quantum algorithm. This requires teamwork; upon successful completion of the module, students should be able to:

- draft and implement basic quantum computing algorithms
- apply quantum computing (simulations) to test these algorithms
- prepare and give oral presentations about their work in front of an audience

**Contents**

Quantum gate algorithms such as Deutsch-Jozsa, Bernstein-Vazirani, Simon, Shor, Grover; phase kick-back, amplitude amplification; swap tests; Hamiltonian simulation, Trotterization, variational quantum computing for optimization

**Prerequisites**

Required: MA-INF 1107 “Fondations of Quantum Computing”

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td>2</td>
<td>30 T / 45 S</td>
<td>2.5</td>
<td></td>
</tr>
<tr>
<td>Exercises</td>
<td>1</td>
<td>15 T / 60 S</td>
<td>2.5</td>
<td></td>
</tr>
</tbody>
</table>

T = face-to-face teaching  
S = independent study

**Graded exams**

Written exam

**Ungraded coursework (required for admission to the exam)**

Successful exercise participation

**Forms of media**

- lecture slides / lecture notes are made available online
- notebooks with programming examples are made available online

**Literature**


MA-INF 1225  Lab Exploring HPC technologies

Workload | Credit points | Duration | Frequency
--- | --- | --- | ---
270 h | 9 CP | 1 semester | every year

Module coordinator
Prof. Dr. Estela Suarez

Lecturer(s)
Prof. Dr. Estela Suarez

Programme
M. Sc. Computer Science

Mode
Optional

Semester
2. or 3.

Learning goals: technical skills
Understanding a use case from complex code developed. Adapting and running applications to different kinds of processing units, taking into account their specific architecture characteristic and programming environments. Understanding and using parallel programming paradigms and high-level programming languages. Designing and executing a benchmarking campaign. Using performance analysis tools, understanding performance bottlenecks and measures to improve them. Software development skills and standards.

Learning goals: soft skills
Collaborating and interacting with application developers, tools developers, and system administrators in a solution oriented manner, taking into account their different “work language” and expertise. Presenting performed work and results obtained and classifying own results into the state-of-the-art. Preparing software documentation.

Contents
The students carry out a practical task (project) in High Performance Computing (HPC), including test of different hardware architectures and software tools, documentation of the implemented software/system. Contents: HPC systems: access/use of compute resources at Jülich Supercomputing Centre; Use of different processor architectures; Software environment, performance analysis tools; Parallel programming; Benchmarking tools/procedures; Performance of applications and scaling behavior, strategies for improvement

Prerequisites
Required:
- Passed the exam of MA-INF 1106 or MA-INF 1108.
- Knowledge modern programming languages (C/C++, Python).
- Willing to stay for at least 2 days per week during 4 weeks at the Jülich Supercomputing Centre, dates to be discussed.

Remarks
Registration first via direct mail communication with the lecturer, in order to identify suitable dates for the stay at JSC.

Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lab</td>
<td>2</td>
<td>4</td>
<td>60 T / 210 S</td>
<td>9</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

Graded exams
Oral presentation, written report

Ungraded coursework (required for admission to the exam)

Forms of media
Own laptop to connect and program on the supercomputers.
<table>
<thead>
<tr>
<th>Literature</th>
</tr>
</thead>
<tbody>
<tr>
<td>• Message Passing Interface Forum: MPI: A Message-Passing Interface Standard, Version 3.1</td>
</tr>
<tr>
<td>• OpenMP Application Programming Interface, Version 4.5, November 2015</td>
</tr>
</tbody>
</table>
MA-INF 1301  Algorithmic Game Theory

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>every 2 years</td>
</tr>
</tbody>
</table>

Module coordinator
Prof. Dr. Thomas Kesselheim

Lecturer(s)
Prof. Dr. Thomas Kesselheim, Senior Prof. Dr. Marek Karpinski

Programme
M. Sc. Computer Science

Mode
Optional

Semester
2. or 3.

Learning goals: technical skills
Knowledge of fundamental results in (algorithmic) game theory and (algorithmic) mechanism design. Techniques and methods related to mathematical modeling of strategic agents. Analyzing and designing systems of strategic agents, with a focus on computational efficiency and performance guarantees.

Learning goals: soft skills
Presentation of solutions and methods, critical discussion of applied methods and techniques

Contents
- basic game theory
- computability and hardness of equilibria
- convergence of dynamics of selfish agents
- (bounds on the) loss of performance due to selfish behavior
- designing incentive-compatible auctions
- maximizing revenue
- designing mechanisms for stable and fair allocations without money

Prerequisites
Recommended:
Introductory knowledge of foundations of algorithms and complexity theory is essential.

Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td>4</td>
<td>60 T / 105 S</td>
<td>5.5</td>
<td></td>
</tr>
<tr>
<td>Exercises</td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
<td></td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

Graded exams
Written exam

Ungraded coursework (required for admission to the exam)
The completion of regularly provided exercise sheets. Each student must present a solution to an exercise in the exercise sessions once.

Literature
- A. Karlin, Y. Peres, Game Theory, Alive, AMS, 2017
MA-INF 1304  Seminar Computational Geometry

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>120 h</td>
<td>4 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

Module coordinator  Prof. Dr. Anne Driemel
Lecturer(s)         Prof. Dr. Anne Driemel, PD Dr. Elmar Langetepe, Dr. Herman Haverkort

Programme M. Sc. Computer Science
Mode             Optional
Semester         2-4

Learning goals: technical skills
To independently study problems at research level, based on research publications, to prepare a concise summary, to present the summary in a scientific talk, to lead a critical discussion with other seminar participants.

Learning goals: soft skills

Contents
Current topics in computational geometry.

Prerequisites
Recommended:
BA-INF 114 – Grundlagen der algorithmischen Geometrie
MA-INF 1203 – Discrete and Computational Geometry

Course meetings
<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Seminar</td>
<td>10</td>
<td>2</td>
<td>30 T / 90 S</td>
<td>4</td>
</tr>
</tbody>
</table>

Graded exams
Oral presentation, written report

Ungraded coursework (required for admission to the exam)

Forms of media
Multimedia projector, black board.

Literature
The relevant literature will be announced.
MA-INF 1305  Graduate Seminar on Applied Combinatorial Optimization

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>180 h</td>
<td>6 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

**Module coordinator**

Prof. Dr. Jens Vygen

**Lecturer(s)**

All lecturers of Discrete Mathematics

**Programme**

M. Sc. Computer Science

**Mode**

Optional

**Semester**

3.

**Learning goals: technical skills**

Competence to understand new theoretical results and practical solutions in VLSI design and related applications, as well as presentation of such results

**Learning goals: soft skills**

Ability to read and understand research papers, abstract thinking, presentation of mathematical results in a talk

**Contents**

Current topics in chip design and related applications

**Prerequisites**

Recommended:

At least 1 of the following:

- MA-INF 1102 – Combinatorial Optimization
- MA-INF 1202 – Chip Design

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Seminar</td>
<td>10</td>
<td>4</td>
<td>60 T / 120 S</td>
<td>6</td>
</tr>
</tbody>
</table>

T = face-to-face teaching

S = independent study

**Graded exams**

Oral presentation, written report

**Ungraded coursework (required for admission to the exam)**

**Literature**

The topics and the relevant literature will be announced towards the end of the previous semester
MA-INF 1307  Seminar Advanced Algorithms

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>120 h</td>
<td>4 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

Module coordinator
Prof. Dr. Thomas Kesselheim

Lecturer(s)
Prof. Dr. Anne Driemel, Prof. Dr. Thomas Kesselheim,
Prof. Dr. Heiko Röglin, PD Dr. Elmar Langetepe, Dr. Herman Haverkort

Programme
M. Sc. Computer Science

Mode
Optional

Semester
3.

Learning goals: technical skills
Presentation of selected advanced topics in algorithm design and various applications

Learning goals: soft skills
Ability to perform individual literature search, critical reading, understanding, and clear didactic presentation

Contents
Advanced topics in algorithm design based on newest research literature

Prerequisites
none

Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Seminar</td>
<td>10</td>
<td>2</td>
<td>30 T / 90 S</td>
<td>4</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

Graded exams
Oral presentation, written report

Ungraded coursework (required for admission to the exam)

Literature
The relevant literature will be announced in time.
MA-INF 1308 Lab Algorithms for Chip Design

### Workload
270 h  
9 CP  
1 semester  
every year

### Module coordinator
Prof. Dr. Jens Vygen

### Lecturer(s)
All lecturers of Discrete Mathematics

### Programme
M. Sc. Computer Science

### Mode
Optional

### Semester
3.

### Learning goals: technical skills
Competence to implement algorithms for VLSI design, efficient handling of very large instances, testing, documentation. Advanced software techniques.

### Learning goals: soft skills
Efficient implementation of complex algorithms, abstract thinking, modelling of optimization problem in VLSI design, documentation of source code

### Contents
A currently challenging problem will be chosen each semester. The precise task will be explained in a meeting in the previous semester.

### Prerequisites
**Recommended:**
At least 3 of the following:
- MA-INF 1102 – Combinatorial Optimization
- MA-INF 1202 – Chip Design
- MA-INF 1205 – Graduate Seminar Discrete Optimization

### Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lab</td>
<td>8</td>
<td>4</td>
<td>60 T / 210 S</td>
<td>9</td>
</tr>
</tbody>
</table>

T = face-to-face teaching  
S = independent study

### Graded exams
Oral presentation, written report

### Ungraded coursework (required for admission to the exam)

### Literature
The topics and the relevant literature will be announced towards the end of the previous semester
MA-INF 1309  Lab Efficient Algorithms: Design, Analysis and Implementation

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>at least every year</td>
</tr>
</tbody>
</table>

**Module coordinator**
Prof. Dr. Heiko Rögl

**Lecturer(s)**
Prof. Dr. Anne Drämel, Prof. Dr. Thomas Kesselheim,
Prof. Dr. Heiko Rögl, PD Dr. Elmar Langetepe, Dr. Herman Haverkort

**Programme**
M. Sc. Computer Science

**Mode**
Optional

**Semester**
3.

**Learning goals: technical skills**
Ability to design, analyze and implement efficient algorithms for selected computational problems.

**Learning goals: soft skills**
ability to work on advanced algorithmic implementation projects, to work in small teams, clear didactic presentation and critical discussion of results

**Contents**
Design of efficient exact and approximate algorithms and data structures for selected computational problems.

**Prerequisites**
none

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lab</td>
<td>8</td>
<td>4</td>
<td>60 T / 210 S</td>
<td>9</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

**Graded exams**
Oral presentation, written report

**Ungraded coursework (required for admission to the exam)**

**Literature**
The relevant literature will be announced in time.
MA-INF 1314  Online Motion Planning

Workload | Credit points | Duration | Frequency
---|---|---|---
270 h | 9 CP | 1 semester | every year

Module coordinator | Lecturer(s)
---|---
PD Dr. Elmar Langetepe | Prof. Dr. Rolf Klein, PD Dr. Elmar Langetepe

Programme | Mode | Semester
---|---|---

Learning goals: technical skills
To acquire fundamental knowledge on topics and methods in online motion planning

Learning goals: soft skills

Contents
Search and exploration in unknown environments (e.g., graphs, cellular environments, polygons, streets), online algorithms, competitive analysis, competitive complexity, functional optimization, shortest watchman route, tethered robots, marker algorithms, spiral search, approximation of optimal search paths.

Prerequisites
Recommended:
BA-INF 114 – Grundlagen der algorithmischen Geometrie

Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td>4</td>
<td>60 T / 105 S</td>
<td>5.5</td>
<td></td>
</tr>
<tr>
<td>Exercises</td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
<td></td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

Graded exams

Oral exam

Ungraded coursework (required for admission to the exam)
The completion of regularly provided exercise sheets. The work can be done in groups of up to three students. A total of 25% of the points must be achieved. Each student must present a solution to an exercise in the exercise sessions once.

Forms of media
Java applets of geometry lab

Literature
Scientific research articles will be recommended in the lecture.
MA-INF 1315  Lab Computational Geometry

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Module coordinator</th>
<th>Lecturer(s)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Prof. Dr. Anne Driemel</td>
<td>Prof. Dr. Anne Driemel, PD Dr. Elmar Langetepe, Dr. Herman Haverkort</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Programme</th>
<th>Mode</th>
<th>Semester</th>
</tr>
</thead>
<tbody>
<tr>
<td>M. Sc. Computer Science</td>
<td>Optional</td>
<td>2.</td>
</tr>
</tbody>
</table>

**Learning goals: technical skills**

Ability to design, analyze, implement and document efficient algorithms for selected problems in computational geometry.

**Learning goals: soft skills**

Ability to properly present, defend and discuss design and implementation decisions, to document software according to given rules and to collaborate with other students in small groups.

**Contents**

Various problems in computational geometry.

**Prerequisites**

none

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lab</td>
<td>8</td>
<td>4</td>
<td>60 T / 210 S</td>
<td>9</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

**Graded exams**

Oral presentation, written report

**Ungraded coursework (required for admission to the exam)**

**Literature**

The relevant literature will be announced in time.
MA-INF 1316  Lab Cryptography

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

Module coordinator

Dr. Michael Nüsken

Lecturer(s)

Dr. Michael Nüsken

Programme

M. Sc. Computer Science

Mode

Optional

Semester

2. or 3.

Learning goals: technical skills

The students will carry out a practical task (project) in the context of Cryptography, including test and documentation of the implemented software/system.

Learning goals: soft skills

Ability to properly present and defend design decisions, to prepare readable documentation of software; skills in constructively collaborating with others in small teams over a longer period of time; ability to classify ones own results into the state-of-the-art of the resp. area

Contents

Prerequisites

Recommended:
Basic knowledge in cryptography is highly recommended, eg. by MA-INF 1103 - Cryptography, MA-INF 1223 - PETs, MA-INF 1209 - Seminar Advanced Topics in Cryptography.

Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lab</td>
<td>8</td>
<td>4</td>
<td>60 T / 210 S</td>
<td>9</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

Graded exams

Oral presentation, written report

Ungraded coursework (required for admission to the exam)
MA-INF 1320  Lab Advanced Algorithms

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>at least every 2 years</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Module coordinator</th>
<th>Lecturer(s)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Prof. Dr. Thomas Kesselheim</td>
<td>Prof. Dr. Thomas Kesselheim, Prof. Dr. Heiko Röglin</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Programme</th>
<th>Mode</th>
<th>Semester</th>
</tr>
</thead>
<tbody>
<tr>
<td>M. Sc. Computer Science</td>
<td>Optional</td>
<td>2. or 3.</td>
</tr>
</tbody>
</table>

**Learning goals: technical skills**
Implementation of algorithms from advanced algorithmic theory, evaluating these algorithm on suitably chosen instances, and discussing how theoretical results transfer to practice.

**Learning goals: soft skills**
Ability to properly present, defend and discuss design and implementation decisions and observed conclusions, and to collaborate with other students in small groups.

**Contents**
Various problems from current research and courses on algorithmic theory.

**Prerequisites**
none

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lab</td>
<td>8</td>
<td>4</td>
<td>60 T / 210 S</td>
<td>9</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

**Graded exams**
Oral presentation, written report

**Ungraded coursework (required for admission to the exam)**

**Literature**
The relevant literature will be announced in time.
MA-INF 1321  Binary Linear and Quadratic Optimization

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>180 h</td>
<td>6 CP</td>
<td>1 semester</td>
<td>at least every 2 years</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Module coordinator</th>
<th>Lecturer(s)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Dr. Sven Mallach</td>
<td>Dr. Sven Mallach</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Programme</th>
<th>Mode</th>
<th>Semester</th>
</tr>
</thead>
<tbody>
<tr>
<td>M. Sc. Computer Science</td>
<td>Optional</td>
<td>2. or 3.</td>
</tr>
</tbody>
</table>

**Learning goals: technical skills**

Deeper understanding of computational methods to solve potentially large-scale mixed-integer programs in practice. Application-specific modelling and reformulation of combinatorial optimization problems, handling quadratic objective functions, algorithm design.

**Learning goals: soft skills**

Social, methodological, and analytical competences via communication, own development, presentation, and critical assessment of problem formulations, algorithms, and solutions covered in the course or the exercises. Learning to abstract, but also learning the limitations of abstraction.

**Contents**

Computational methods in (mixed-)integer programming such as cutting plane separation and branch-and-bound along with a short and accessible introduction into their theoretical basis. Study of practically relevant binary linear and binary quadratic optimization problems, e.g., Maximum Cut, Linear Ordering and variants of the Traveling Salesman problem, along with the particular separation problems arising there. If there is time, linearizations of quadratic objective functions and more sophisticated formulations of binary quadratic problems are discussed.

**Prerequisites**

none

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td></td>
<td>2</td>
<td>30 T / 45 S</td>
<td>2.5</td>
</tr>
<tr>
<td>Exercises</td>
<td></td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
</tr>
</tbody>
</table>

T = face-to-face teaching  
S = independent study

**Graded exams**

Oral exam

**Ungraded coursework (required for admission to the exam)**

Successful exercise participation
MA-INF 1322  Seminar Focus Topics in High Performance Computing

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>120 h</td>
<td>4 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

**Module coordinator**
Prof. Dr. Estela Suarez

**Lecturer(s)**
Prof. Dr. Estela Suarez

**Programme**
M. Sc. Computer Science

**Mode**
Optional

**Semester**
2. or 3.

**Learning goals: technical skills**
Ability to perform individual literature search, critical reading, understanding, prepare a concise summary, and clear didactic presentation

**Learning goals: soft skills**
Ability to present and critically discuss these results in the framework of the corresponding area

**Contents**
General topics and trends in high performance computing, based on recent review and research literature

**Prerequisites**
Recommended:
Interest in High Performance Computing

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Seminar</td>
<td>10</td>
<td>2</td>
<td>30 T / 90 S</td>
<td>4</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

**Graded exams**
Oral presentation, written report

**Ungraded coursework (required for admission to the exam)**

**Literature**
Literature and further information about this seminar will be announced in time in the website of lecturer.
MA-INF 1323  Computational Topology

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>at least every 2 years</td>
</tr>
</tbody>
</table>

**Module coordinator**
Prof. Dr. Anne Driemel

**Lecturer(s)**
Prof. Dr. Anne Driemel, Dr. Benedikt Kolbe

**Programme**
M. Sc. Computer Science

**Mode**
Optional

**Semester**
2. or 3.

**Learning goals: technical skills**
Knowledge of fundamental theorems and concepts in the area of computational topology in particular, persistent homology and topological data analysis; design and analysis of combinatorial algorithms in topological contexts; analysis of the complexity; to apply this knowledge autonomously to solving new problems and analysing new data sets.

**Learning goals: soft skills**
Social competence (communication, presenting one’s own solutions, goal-oriented discussions in teams), methodical competence (analysis, abstraction, proofs), individual competence (commitment and willingness to learn, creativity, perseverance).

**Contents**
Fundamental concepts of relative homology and cohomology theory and persistence theory in computational settings, category theory in this context, algorithms for the computation of (persistent) homology, (extended) persistence modules and their decompositions, Morse theory, duality theorems, quiver representation theory, stability of persistence diagrams and barcodes, algebraic stability, topological filtrations, multiparameter persistence, invariants of persistence, topological data analysis, applications to shape pattern recognition, machine learning, identification of geometric objects.

**Prerequisites**
none

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td></td>
<td>4</td>
<td>60 T / 105 S</td>
<td>5.5</td>
</tr>
<tr>
<td>Exercises</td>
<td></td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

**Graded exams**
Written exam

**Ungraded coursework (required for admission to the exam)**
Successful exercise participation

**Literature**
2 Graphics, Vision, Audio

MA-INF 2113 L2E2 6 CP Foundations of Audio Signal Processing .................................. 41
MA-INF 2201 L4E2 9 CP Computer Vision .......................................................... 42
MA-INF 2202 L4E2 9 CP Computer Animation .................................................. 43
MA-INF 2203 L4E2 9 CP Selected Topics in Signal Processing .................................. 44
MA-INF 2206 Sem2 4 CP Seminar Vision .......................................................... 45
MA-INF 2207 Sem2 4 CP Seminar Graphics ....................................................... 46
MA-INF 2208 Sem2 4 CP Seminar Audio ........................................................... 47
MA-INF 2209 L4E2 9 CP Advanced Topics in Computer Graphics I ......................... 48
MA-INF 2212 L2E2 6 CP Pattern Matching and Machine Learning for Audio Signal Processing 50
MA-INF 2213 L3E1 6 CP Advanced Computer Vision .......................................... 51
MA-INF 2214 L2E2 6 CP Computational Photography ......................................... 52
MA-INF 2215 Sem2 4 CP Seminar Digital Material Appearance ............................. 53
MA-INF 2216 Lab4 9 CP Lab Visual Computing .................................................. 54
MA-INF 2217 L2E2 6 CP Advanced Deep Learning for Graphics .............................. 55
MA-INF 2218 L2E2 6 CP Video Analytics ......................................................... 56
MA-INF 2219 Sem2 4 CP Seminar Visualization and Medical Image Analysis ............ 57
MA-INF 2220 Lab4 9 CP Lab Visualization and Medical Image Analysis .................... 58
MA-INF 2221 Sem2 4 CP Seminar Visual Computing ........................................... 59
MA-INF 2222 L4E2 9 CP Visual Data Analysis ................................................... 60
MA-INF 2223 Sem2 4 CP Seminar Advances in Multimodal Learning ....................... 61
MA-INF 2224 Lab4 9 CP Lab Challenges in Computer Vision .................................. 62
MA-INF 2225 L2E2 6 CP Discrete Models for Visual Computing ............................. 63
MA-INF 2226 Lab4 9 CP Lab Geometry Processing ............................................ 64
MA-INF 2227 Lab4 9 CP Lab 3D Animation ....................................................... 65
MA-INF 2230 Lab4 9 CP Lab Vision ................................................................. 66
MA-INF 2231 Lab4 9 CP Lab Graphics .............................................................. 67
MA-INF 2308 Lab4 9 CP Lab Audio ................................................................. 68
MA-INF 2310 L4E2 9 CP Advanced Topics in Computer Graphics II ......................... 69
MA-INF 2312 L3E1 6 CP Image Acquisition and Analysis in Neuroscience ............... 71
MA-INF 2313 L2E2 6 CP Deep Learning for Visual Recognition ............................. 72
MA-INF 2315 Sem2 4 CP Seminar Computational Photography ............................. 73
MA-INF 2316 Lab4 9 CP Lab Digital Material Appearance .................................... 74
MA-INF 2113  Foundations of Audio Signal Processing

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>180 h</td>
<td>6 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

Module coordinator
apl. Prof. Dr. Frank Kurth

Lecturer(s)
apl. Prof. Dr. Frank Kurth, Prof. Dr. Michael Clausen

Programme
M. Sc. Computer Science

Mode
Optional

Semester
1.

Learning goals: technical skills
- Introduction to basic concepts of analog and digital signal processing;
- Applications in the field of Audio Signal Processing;
- Signal Processing Algorithms;
- Implementing basic Signal Processing Algorithms

Learning goals: soft skills
Solving basic Signal Processing Problems; Implementing Signal Processing Algorithms using state-of-the-art software frameworks; Capability to analyze; Time management; Presentation skills; Discussing own solutions and solutions of others, and working in groups.

Contents
Theoretical introduction to analog and digital Signal Processing; Fourier Transforms; Analog to digital Conversion; Digital Filters; Audio Signal Processing Applications; Filter banks; Windowed Fourier Transform; 2D-Signal Processing

Prerequisites
none

Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload [h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td>2</td>
<td>30 T / 45 S</td>
<td>2.5</td>
<td></td>
</tr>
<tr>
<td>Exercises</td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
<td></td>
</tr>
</tbody>
</table>

Graded exams
Written exam

Ungraded coursework (required for admission to the exam)
Successful exercise participation

Forms of media
Slides, Blackboard, Whiteboard
MA-INF 2201  Computer Vision

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

Module coordinator
Prof. Dr. Jürgen Gall

Lecturer(s)
Prof. Dr. Jürgen Gall

Programme
M. Sc. Computer Science

Mode
Optional

Semester
1. or 2.

Learning goals: technical skills
Students will learn about various mathematical methods and their applications to computer vision problems.

Learning goals: soft skills
Productive work in small teams, development and realization of individual approaches and solutions, critical reflection of competing methods, discussion in groups.

Contents
The class will cover a number of mathematical methods and their applications in computer vision. For example, linear filters, edges, derivatives, Hough transform, segmentation, graph cuts, mean shift, active contours, level sets, MRFs, expectation maximization, background subtraction, temporal filtering, active appearance models, shapes, optical flow, 2d tracking, cameras, 2d/3d features, stereo, 3d reconstruction, 3d pose estimation, articulated pose estimation, deformable meshes, RGBD vision.

Prerequisites
Recommended:
Basic knowledge of linear algebra, analysis, probability theory, Python programming

Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td>4</td>
<td>60 T / 105 S</td>
<td>5.5</td>
<td></td>
</tr>
<tr>
<td>Exercises</td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
<td></td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

Graded exams
Written exam

Ungraded coursework (required for admission to the exam)
Successful exercise participation

Literature
- R. Hartley, A. Zisserman: Multiple View Geometry in Computer Vision
- R. Szeliski: Computer Vision: Algorithms and Applications
- S. Prince: Computer Vision: Models, Learning, and Inference
MA-INF 2202 Computer Animation

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

Module coordinator
Prof. Dr. Björn Krüger

Lecturer(s)
Prof. Dr. Björn Krüger

Programme
M. Sc. Computer Science

Mode
Optional

Semester
2.

Learning goals: technical skills
Students will learn fundamental paradigms used in computer animation. They will learn to use mathematical models of motions to come up with algorithmic solutions of problems of the synthesis of motions of virtual characters.

Learning goals: soft skills
Social competences (work in groups), communicative skills (written and oral presentation)

Contents
Fundamentals of computer animation; kinematics; representations of motions; motion capturing; motion editing; motion synthesis; facial animations

Prerequisites
none

Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td>4</td>
<td>60 T / 105 S</td>
<td>5.5</td>
<td></td>
</tr>
<tr>
<td>Exercises</td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
<td></td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

Graded exams
Written exam

Ungraded coursework (required for admission to the exam)
The completion of regularly provided exercise sheets. The work can be done in groups of up to three students. A total of 50% of the points must be achieved.

Literature
MA-INF 2203  Selected Topics in Signal Processing

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

**Module coordinator**  
apl. Prof. Dr. Frank Kurth

**Lecturer(s)**  
apl. Prof. Dr. Frank Kurth, Prof. Dr. Michael Clausen

**Programme**  
M. Sc. Computer Science

**Mode**  
Optional

**Semester**  
2.

**Learning goals: technical skills**
Learning advanced as well as state of the art topics and techniques in digital signal processing. Study examples from the field of digital audio signal processing with a focus on music audio. Develop skills for analysing audio signals and designing audio features for selected application scenarios. Mathematical modelling of signal processing problems in practical applications. Design and implementation of corresponding algorithms and data structures solving those problems. Efficiency issues.

**Learning goals: soft skills**

**Contents**
Advanced techniques for filter design, design and extraction of features describing multimedia signals, efficient DSP algorithms, general concepts for content-based analysis of multimedia signals. Selected signal processing applications, for example content-based music analysis, signal compression, denoising, source separation.

**Prerequisites**
none

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td>4</td>
<td>60 T / 105 S</td>
<td>5.5</td>
<td></td>
</tr>
<tr>
<td>Exercises</td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
<td></td>
</tr>
</tbody>
</table>

T = face-to-face teaching  
S = independent study

**Graded exams**
Written exam

**Ungraded coursework (required for admission to the exam)**
Successful exercise participation

**Literature**
- Lecture script and selected research publications
MA-INF 2206 Seminar Vision

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>120 h</td>
<td>4 CP</td>
<td>1 semester</td>
<td>every semester</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Module coordinator</th>
<th>Lecturer(s)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Prof. Dr. Jürgen Gall</td>
<td>Prof. Dr. Jürgen Gall</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Programme</th>
<th>Mode</th>
<th>Semester</th>
</tr>
</thead>
<tbody>
<tr>
<td>M. Sc. Computer Science</td>
<td>Optional</td>
<td>2. or 3.</td>
</tr>
</tbody>
</table>

**Learning goals: technical skills**
Ability to understand new research results presented in original scientific papers.

**Learning goals: soft skills**
Ability to present and to critically discuss these results in the framework of the corresponding area.

**Contents**
Current conference and journal papers.

**Prerequisites**
Required: MA-INF 2201 – Computer Vision

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Seminar</td>
<td>10</td>
<td>2</td>
<td>30 T / 90 S</td>
<td>4</td>
</tr>
</tbody>
</table>

\[ T = \text{face-to-face teaching}\]

\[ S = \text{independent study}\]

**Graded exams**
Oral presentation, written report

**Ungraded coursework (required for admission to the exam)**
MA-INF 2207  Seminar Graphics

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>120 h</td>
<td>4 CP</td>
<td>1 semester</td>
<td>every semester</td>
</tr>
</tbody>
</table>

**Module coordinator**

Prof. Dr. Reinhard Klein

**Lecturer(s)**

Prof. Dr. Reinhard Klein

**Programme**

M. Sc. Computer Science

**Mode**

Optional

**Semester**

2. or 3.

**Learning goals: technical skills**

Ability to understand new research results presented in original scientific papers.

**Learning goals: soft skills**

Ability to present and to critically discuss these results in the framework of the corresponding area.

**Contents**

Current conference and journal papers.

**Prerequisites**

Recommended:

Mathematical background (multidimensional analysis and linear algebra, basic numerical methods)

Basic knowledge in Computer Graphics

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Seminar</td>
<td>10</td>
<td>2</td>
<td>30 T / 90 S</td>
<td>4</td>
</tr>
</tbody>
</table>

T = face-to-face teaching

S = independent study

**Graded exams**

Oral presentation, written report

**Ungraded coursework (required for admission to the exam)**
MA-INF 2208 Seminar Audio

**Workload** | **Credit points** | **Duration** | **Frequency**
---|---|---|---
120 h | 4 CP | 1 semester | every semester

**Module coordinator**
apl. Prof. Dr. Frank Kurth

**Lecturer(s)**
apl. Prof. Dr. Frank Kurth, Dr. Michael Clausen

**Programme**
M. Sc. Computer Science

**Mode**
Optional

**Semester**
2.

**Learning goals: technical skills**
Ability to understand new research results presented in original scientific papers.

**Learning goals: soft skills**
Ability to present and to critically discuss these results in the framework of the corresponding area.

**Contents**
Current conference and journal papers.

**Prerequisites**
none

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Seminar</td>
<td>10</td>
<td>2</td>
<td>30 T / 90 S</td>
<td>4</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

**Graded exams**
Oral presentation, written report

**Ungraded coursework (required for admission to the exam)**
MA-INF 2209  Advanced Topics in Computer Graphics I

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

**Module coordinator**
Prof. Dr. Reinhard Klein
**Lecturer(s)**
Prof. Dr. Reinhard Klein

**Programme**
M. Sc. Computer Science
**Mode**
Optional
**Semester**
2. or 3.

**Learning goals: technical skills**
Analytical formulation of problems related to rendering. Knowledge of principles, techniques and algorithms to
• recognize and understand the physical quantities of light transport
• explain a range of surface and volumetric material models
• explain the rendering and radiative transfer equations
• design and implement methods to solve these equations, especially Monte Carlo methods
• Assess / Evaluate the performance and conceptual limits of the implemented simulation code

**Learning goals: soft skills**
Based on the knowledge and skills acquired students should be able to
• read and judge current scientific literature in the area of rendering
• identify the major literature concerning a given problem in rendering and gain an overview of the current state of the art
• discuss problems concerning rendering with researchers from different application fields
• present, propose and communicate different solutions and work in a team to solve a rendering problem

**Contents**
This course introduces the basic physical quantities as well as the mathematical and algorithmic tools required to understand and simulate the light interaction with objects and different materials in a 3D scene. We will discuss how to solve the mathematical problem numerically in order to create realistic images. Advanced topics include participating media, material models for sub-surface light transport, and Markov Chain Monte Carlo Methods.
Topics among others will be
• rendering and radiative transfer equation
• methods and algorithms to solve these equations, radiosity, Monte Carlo, photon mapping
• analytical and data driven surface and subsurface material models, especially BRDF, BSSRDF models
• differentiable rendering
In addition, results from state-of-the-art research will be presented.

**Prerequisites**
Recommended: Recommended but not enforced: basic knowledge in computer graphics, (numerical) analysis and linear algebra, C++

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td>4</td>
<td>60 T / 105 S</td>
<td>5.5</td>
<td></td>
</tr>
<tr>
<td>Exercises</td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
<td></td>
</tr>
</tbody>
</table>

**Graded exams**
Oral presentation, written report

**Ungraded coursework (required for admission to the exam)**
The completion of regularly provided exercise sheets. The work can be done in groups of up to two students. A total of 50% of the points must be achieved. For 70% of the exercise sheets, 50% of the points must be achieved for each sheet. The exercises are divided into theoretical and practical exercises, and the points to be achieved apply separately to both categories.
<table>
<thead>
<tr>
<th>Literature</th>
</tr>
</thead>
<tbody>
<tr>
<td>• M. Pharr, W. Jakob, and G. Humphreys, Physically Based Rendering: From Theory to Implementation (3rd edition), 2018</td>
</tr>
</tbody>
</table>
MA-INF 2212  Pattern Matching and Machine Learning for Audio Signal Processing

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>180 h</td>
<td>6 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

Module coordinator

apl. Prof. Dr. Frank Kurth

Lecturer(s)

apl. Prof. Dr. Frank Kurth, Prof. Dr. Michael Clausen

Programme

M. Sc. Computer Science

Mode

Optional

Semester

2.

Learning goals: technical skills

• Introduction into selected topics of digital signal processing;
• Applications in the field of Audio Signal Processing;
• Methods of Automatic Pattern Recognition

Learning goals: soft skills

Audio Signal Processing Applications; Extended programming skills for signal processing applications; Capability to analyze; Time management; Presentation skills; Discussing own solutions and solutions of others, and working in groups.

Contents

The lecture is presented in modular form, where each module is motivated from the application side. The presented topics are: Windowed Fourier transforms; Audio Identifications; Audio Matching; Signal Classification; Hidden Markov Models; Support Vector Machines

Prerequisites

none

Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td>2</td>
<td>30 T / 45 S</td>
<td>2.5</td>
<td></td>
</tr>
<tr>
<td>Exercises</td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
<td></td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

Graded exams

Written exam

Ungraded coursework (required for admission to the exam)

The completion of regularly provided exercise sheets. The work can be done in groups of up to four students. A total of 50% of the points must be achieved.

Forms of media

Slides, Blackboard, Whiteboard
MA-INF 2213  Advanced Computer Vision

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>180 h</td>
<td>6 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

Module coordinator
Prof. Dr. Jürgen Gall

Lecturer(s)
Prof. Dr. Jürgen Gall

Programme
M. Sc. Computer Science

Mode
Optional

Semester
2. or 3.

Learning goals: technical skills
Students will learn about various learning methods and their applications to computer vision problems.

Learning goals: soft skills
Productive work in small teams, development and realization of individual approaches and solutions, critical reflection of competing methods, discussion in groups.

Contents
The class will cover a number of learning methods and their applications in computer vision. For example, linear methods for classification and regression, Gaussian processes, random forests, SVMs and kernels, convolutional neural networks, vision transformer, generative adversarial networks, diffusion models, structured learning, image classification, object detection, action recognition, pose estimation, face analysis, tracking, image synthesis.

Prerequisites
Required:
MA-INF 2201 – Computer Vision

Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td></td>
<td>3</td>
<td>45 T / 45 S</td>
<td>3</td>
</tr>
<tr>
<td>Exercises</td>
<td></td>
<td>1</td>
<td>15 T / 75 S</td>
<td>3</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

Graded exams
Oral exam

Ungraded coursework (required for admission to the exam)
The completion of regularly provided exercise sheets. The work can be done in groups of up to two students. A total of 50% of the points must be achieved.
MA-INF 2214  Computational Photography

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>180 h</td>
<td>6 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

**Module coordinator**
Prof. Dr. Matthias Hullin

**Lecturer(s)**
Prof. Dr. Matthias Hullin

**Programme**
M. Sc. Computer Science

**Mode**
Optional

**Semester**
2. or 3.

**Learning goals: technical skills**

**Learning goals: soft skills**
- to read and understand current literature in the field
- to implement standard computational photography techniques
- to propose and implement solutions to a given problem
- to follow good scientific practice by planning, documenting and communicating their work

**Contents**
- Image sensors
- Optics
- Panoramas
- Light fields
- Signal processing and inverse problems
- Color, perception and HDR
- Reflectance fields and light transport matrices

**Prerequisites**
Required:
Basic knowledge in computer graphics, data structures, multidimensional analysis und linear algebra, numerical analysis and numerical linear algebra, C++ or MATLAB

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td>2</td>
<td>30 T / 45 S</td>
<td>2.5</td>
<td></td>
</tr>
<tr>
<td>Exercises</td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
<td></td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

**Graded exams**
Oral exam

**Ungraded coursework (required for admission to the exam)**
(i) The completion of regularly provided exercise sheets. The work can be done in groups of up to two students. A total of 50% of the points must be achieved. The exercises are divided into theoretical and practical exercises, and the points to be achieved apply separately to both categories. Each student must present a solution to an exercise in the exercise sessions twice. (ii) The completion of a programming project. The work is done in groups of two to four students, depending on the total number of students taking the course. The results of the programming project must be presented in class.
MA-INF 2215  Seminar Digital Material Appearance

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>120 h</td>
<td>4 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

**Module coordinator**
Prof. Dr. Matthias Hullin

**Lecturer(s)**
Prof. Dr. Matthias Hullin

**Programme**
M. Sc. Computer Science

**Mode**
Optional

**Semester**
2.

**Learning goals: technical skills**
Ability to understand new research results presented in original scientific papers.

**Learning goals: soft skills**
Ability to present and to critically discuss these results in the framework of the corresponding area.

**Contents**
Current conference and journal papers

**Prerequisites**
none

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Seminar</td>
<td>10</td>
<td>2</td>
<td>30 T / 90 S</td>
<td>4</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

**Graded exams**
Oral presentation, written report

**Ungraded coursework (required for admission to the exam)**
MA-INF 2216  Lab Visual Computing

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

**Module coordinator**

Prof. Dr. Florian Bernard

**Lecturer(s)**

Prof. Dr. Florian Bernard

**Programme**

M. Sc. Computer Science

**Mode**

Optional

**Semester**

1-3

**Learning goals: technical skills**

Students will carry out a practical task (project) in the context of visual computing, including test and documentation of the implemented software/system.

**Learning goals: soft skills**

Ability to properly present and defend design decisions, to prepare readable documentation of software; skills in constructively collaborating with others in small teams over a longer period of time; ability to classify ones own results into the state-of-the-art of the resp. area.

**Contents**

This lab introduces visual computing methods and applications. You will get a chance to study the methods in depth by implementing them and running experiments. At the end of the semester, you will present the method, give a short demonstration and hand in a report describing the method and experimental outcomes.

**Prerequisites**

none

**Course meetings**

**Teaching format**

<table>
<thead>
<tr>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lab</td>
<td>8</td>
<td>4</td>
<td>60 T / 210 S</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

**Graded exams**

Oral presentation, written report

**Ungraded coursework (required for admission to the exam)**
MA-INF 2217  Advanced Deep Learning for Graphics

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>180 h</td>
<td>6 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

**Module coordinator**
Prof. Dr. Reinhard Klein

**Lecturer(s)**
Dr. Michael Weinmann

**Programme**
M. Sc. Computer Science

**Mode**
Optional

**Semester**
1-4.

**Learning goals: technical skills**
Students will be introduced to adapt and apply deep learning techniques to various applications in computer graphics.

**Learning goals: soft skills**
Productive work in small teams, development and realization of individual approaches and solutions, critical reflection of competing methods, discussion in groups.

**Contents**
This course focuses on cutting-edge Deep Learning techniques for computer graphics. After a brief review of CNNs the focus will be laid on autoencoders, generative models and the extension of these methods to graph- and manifold-structured data. Applications discussed will include inverse problems in computer graphics and the synthesis of models including data completion and super-resolution.

**Prerequisites**
Recommended:
The course will build upon the basics of machine learning as well as fundamentals and basic architectures of neural networks. Therefore, it is highly recommended to have taken Deep Learning for Visual Recognition or a similar course as a prerequisite. Exercises will be a mix of theory and practical (Python).

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td>2</td>
<td>30 T / 45 S</td>
<td>2.5</td>
<td></td>
</tr>
<tr>
<td>Exercises</td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
<td></td>
</tr>
</tbody>
</table>

\[ T = \text{face-to-face teaching} \]
\[ S = \text{independent study} \]

**Graded exams**
Written exam

**Ungraded coursework (required for admission to the exam)**
Successful exercise participation

**Literature**
No required text, supplemental readings will be given in class.
MA-INF 2218  Video Analytics

<table>
<thead>
<tr>
<th><strong>Workload</strong></th>
<th><strong>Credit points</strong></th>
<th><strong>Duration</strong></th>
<th><strong>Frequency</strong></th>
</tr>
</thead>
<tbody>
<tr>
<td>180 h</td>
<td>6 CP</td>
<td>1 semester</td>
<td>at least every 2 years</td>
</tr>
</tbody>
</table>

**Module coordinator**

Prof. Dr. Jürgen Gall

**Lecturer(s)**

Prof. Dr. Jürgen Gall

**Programme**

M. Sc. Computer Science

**Mode**

Optional

**Semester**

2-3.

**Learning goals: technical skills**

Students will learn advanced techniques for analyzing video data.

**Learning goals: soft skills**

Productive work in small teams, development and realization of a state-of-the-art system for video analysis.

**Contents**

The class will discuss state-of-the-art methods for several tasks of video analysis. For example, action recognition, hidden Markov models, 3D convolutional neural networks, temporal convolutional networks, recurrent neural networks, temporal action segmentation, weakly supervised learning, self-supervised learning, anticipation and forecasting.

**Prerequisites**

Required:

MA-INF 2201 – Computer Vision

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload</th>
<th>h</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td></td>
<td>2</td>
<td>30 T / 45 S</td>
<td>2.5</td>
<td></td>
</tr>
<tr>
<td>Exercises</td>
<td></td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
<td></td>
</tr>
</tbody>
</table>

T = face-to-face teaching

S = independent study

**Graded exams**

Oral exam

**Ungraded coursework (required for admission to the exam)**

The completion of regularly provided exercise sheets. The work can be done in groups of up to three students. A total of 50% of the points must be achieved.
MA-INF 2219  Seminar Visualization and Medical Image Analysis

Workload                      Credit points | Duration | Frequency
120 h                          4 CP         | 1 semester | every semester

Module coordinator             Lecturer(s)
Prof. Dr. Thomas Schultz       Prof. Dr. Thomas Schultz

Programme
M. Sc. Computer Science

Mode
Optional

Semester
2.

Learning goals: technical skills
Ability to understand new research results presented in original scientific papers.

Learning goals: soft skills
Ability to present and to critically discuss scientific results in the context of the current state of the art. Ability to perform an independent search for relevant scientific literature.

Contents
Current conference and journal papers

Prerequisites
Recommended:
At least one of the following:
• MA-INF 2222 – Visual Data Analysis
• MA-INF 2312 – Image Acquisition and Analysis in Neuroscience

Course meetings
Teaching format | Group size | h/week | Workload[h] | CP | T = face-to-face teaching |
Seminar         | 10         | 2      | 30 T / 90 S | 4  | S = independent study |

Graded exams
Oral presentation, written report

Ungraded coursework (required for admission to the exam)
MA-INF 2220  Lab Visualization and Medical Image Analysis

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>every semester</td>
</tr>
</tbody>
</table>

Module coordinator
Prof. Dr. Thomas Schultz

Lecturer(s)
Prof. Dr. Thomas Schultz

Programme
M. Sc. Computer Science

Mode
Optional

Semester
2.

Learning goals: technical skills
The students will carry out a practical task (project) in the context of data visualization and visual analytics or medical image analysis, including test and documentation of the implemented software/system.

Learning goals: soft skills
Ability to properly present and defend design decisions, to prepare readable documentation of software; skills in constructively collaborating with others in small teams over a longer period of time; ability to classify one's own results into the state-of-the-art of the resp. area.

Contents

Prerequisites

Recommended:
At least one of the following:
- MA-INF 2222 – Visual Data Analysis
- MA-INF 2312 – Image Acquisition and Analysis in Neuroscience

Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lab</td>
<td>8</td>
<td>4</td>
<td>60 T / 210 S</td>
<td>9</td>
</tr>
</tbody>
</table>

T = face-to-face teaching

S = independent study

Graded exams
Oral presentation, written report

Ungraded coursework (required for admission to the exam)
MA-INF 2221  Seminar Visual Computing

**Workload**
120 h  
**Credit points**
4 CP  
**Duration**
1 semester  
**Frequency**
at least every year

**Module coordinator**
Prof. Dr. Florian Bernard

**Lecturer(s)**
Prof. Dr. Florian Bernard

**Programme**
M. Sc. Computer Science

**Mode**
Optional

**Semester**
2. or 3.

### Learning goals: technical skills
Ability to understand new research results presented in original scientific papers.

### Learning goals: soft skills
Ability to present and to critically discuss these results in the framework of the corresponding area.

### Contents
Current conference and journal papers.

### Prerequisites
**Required:**
No formal requirements. Participants are expected to have some previous exposure to at least one of the following:
- visual computing (e.g. computer vision, computer graphics, 3D shape analysis, image analysis, etc.),
- mathematical optimisation (e.g. combinatorial/continuous, convex/non-convex, etc.), or
- machine learning.

### Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
<th>T = face-to-face teaching</th>
<th>S = independent study</th>
</tr>
</thead>
<tbody>
<tr>
<td>Seminar</td>
<td>10</td>
<td>2</td>
<td>30 T / 90 S</td>
<td>4</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

### Graded exams
Oral presentation, written report

### Ungraded coursework (required for admission to the exam)
MA-INF 2222  Visual Data Analysis

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

**Module coordinator**
Prof. Dr. Thomas Schultz

**Lecturer(s)**
Prof. Dr. Thomas Schultz

**Programme**
M. Sc. Computer Science

**Mode**
Optional

**Semester**
1-4.

**Learning goals: technical skills**
Ability to design, implement, and make proper use of systems for visual data analysis. Knowledge of algorithms and techniques for the visualization of multi-dimensional data, graphs, as well as scalar, vector, and tensor fields.

**Learning goals: soft skills**
Productive work in small teams, self-dependent solution of practical problems in the area of visual data analysis, critical reflection on visualization design, presentation of solution strategies and implementations, self management

**Contents**
This class provides a broad overview of principles and algorithms for data analysis via interactive visualization. Specific topics include perceptual principles, luminance and color, visualization analysis and design, integration of visual with statistical data analysis and machine learning, as well as specific algorithms and techniques for the display of multidimensional data, dimensionality reduction, graphs, direct and indirect volume visualization, vector field and flow visualization, as well as tensor field visualization.

**Prerequisites**
Recommended:
Students are recommended to have a basic knowledge in linear algebra and calculus, as well as proficiency in programming.

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[CP]</th>
<th>T = face-to-face teaching</th>
<th>S = independent study</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td>4</td>
<td>4</td>
<td>60 T / 105 S</td>
<td>5.5</td>
<td></td>
</tr>
<tr>
<td>Exercises</td>
<td>2</td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
<td></td>
</tr>
</tbody>
</table>

**Graded exams**
Written exam

**Ungraded coursework (required for admission to the exam)**
The completion of regularly provided exercise sheets. The work can be done in groups of up to three students. A total of 50% of the points must be achieved. Each student must present a solution to an exercise in the exercise sessions once.

**Literature**
M. Ward et al., Interactive Data Visualization: Foundations, Techniques, and Applications. CRC Press, 2010
MA-INF 2223  Seminar Advances in Multimodal Learning

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>120 h</td>
<td>4 CP</td>
<td>1 semester</td>
<td>every semester</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Module coordinator</th>
<th>Lecturer(s)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Prof. Dr. Hildegard Kühne</td>
<td>Prof. Dr. Hildegard Kühne</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Programme</th>
<th>Mode</th>
<th>Semester</th>
</tr>
</thead>
<tbody>
<tr>
<td>M. Sc. Computer Science</td>
<td>Optional</td>
<td>2</td>
</tr>
</tbody>
</table>

**Learning goals: technical skills**

Presentation of selected advanced topics in computer vision and multimodal learning and various applications

**Learning goals: soft skills**

Ability to perform individual literature search, critical reading, understanding, and clear didactic presentation

**Contents**

This seminar will cover most recent advancements and publications in multimodal learning, which is the integration of multiple data sources or multiple modalities for more complex machine learning applications. This can also include reviews of emerging techniques, including unsupervised approaches, deep learning, transfer learning, and reinforcement learning to combine multiple modalities such as images, audio, video, joint feature learning, and natural language processing. It can further cover techniques for data fusion and the role they play in successful applications of multimodal learning. Students will have an opportunity to evaluate and experiment with public code if available. Goal is to develop a better understanding of the possibilities and challenges of multimodal learning.

**Prerequisites**

Required:
none

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Seminar</td>
<td>10</td>
<td>2</td>
<td>30 T / 90 S</td>
<td>4</td>
</tr>
</tbody>
</table>

T = face-to-face teaching  
S = independent study

**Graded exams**

Oral presentation, written report

**Ungraded coursework (required for admission to the exam)**

Successful exercise participation

**Literature**

The relevant literature will be announced in time.
MA-INF 2224  Lab Challenges in Computer Vision

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>every semester</td>
</tr>
</tbody>
</table>

Module coordinator
Prof. Dr. Hildegard Kühne

Lecturer(s)
Prof. Dr. Hildegard Kühne

Programme
M. Sc. Computer Science

Mode
Optional

Semester
2.

Learning goals: technical skills
Students will carry out a practical task (project) in the context of computer vision and/or multimodal learning, including evaluation and documentation of the implemented software/system.

Learning goals: soft skills
Ability to implement and evaluate a scientific approach; ability to classify ones own results into the state-of-the-art of the resp. area; skills in constructively collaborating with others in small teams over a longer period of time.

Contents
This Programming Project focuses on exploring the challenges in modern Computer Vision algorithms and model development. The project will track the latest progress in the field and the associated challenges in different application areas, such as video understanding as well as general computer vision topics. The project will include a hands-on implementation of various techniques in current computer vision systems to identify and resolve problems, and to evaluate results in comparison to public benchmarks. It will further provide an understanding of the characteristics of models and benchmarks such as generalization and robustness. The project should provide insights on the development of novel computer vision technology in response to upcoming challenges.

Prerequisites
Required:
Practical experience in deep learning

Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lab</td>
<td>8</td>
<td>4</td>
<td>60 T / 210 S</td>
<td>9</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

Graded exams
Oral presentation, written report

Ungraded coursework (required for admission to the exam)
Successful exercise participation
MA-INF 2225  Discrete Models for Visual Computing

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>180 h</td>
<td>6 CP</td>
<td>1 semester</td>
<td>at least every 2 years</td>
</tr>
</tbody>
</table>

Module coordinator
Prof. Dr. Florian Bernard

Programme
M. Sc. Computer Science

Mode
Optional

Semester
2. or 4.

Learning goals: technical skills
- Ability to implement basic visual computing algorithms, understanding their strengths and shortcomings
- Mathematical modelling of computational problems in visual computing
- Gain an intuition which algorithm is best applied for which problem in visual computing, so that practical problems in these areas can be solved

Learning goals: soft skills
- Problem solving skills: ability to identify and utilise analogies between new problems and previously seen ones
- Analytical and abstract thinking: develop a general intuition of computational problems, being able to adopt different perspectives of particular concepts

Contents
This module focuses on discrete models that frequently occur in the field of visual computing (VC).
In addition to algorithms, this module will also cover modelling aspects that are relevant for solving practical problems in VC.
The contents include:
- Graph-based models (e.g. linear and quadratic assignment, network flows, product graph formalisms, dynamic programming and their application)
- Continuous algorithms for discrete problems (e.g. convex & spectral relaxations, projection methods, path-following and their application)
- Deep Learning for discrete domains (e.g. differentiable programming, graph neural networks, deep learning on meshes)

Prerequisites
Required:
No formal prerequisites
Recommended:
Participants are expected to have a high level of mathematical maturity (in particular, a good working knowledge of linear algebra and calculus is essential). A basic understanding of graph theory is advantageous.

Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td>2</td>
<td>30 T / 45 S</td>
<td>2.5</td>
<td></td>
</tr>
<tr>
<td>Exercises</td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
<td></td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

Graded exams
Oral exam

Ungraded coursework (required for admission to the exam)
The completion of regularly provided exercise sheets. The work can be done in groups of up to two students. A total of 50% of the points must be achieved.
**MA-INF 2226  Lab Geometry Processing**

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>at least every 2 years</td>
</tr>
</tbody>
</table>

**Module coordinator**

Lecturer(s)

Jun. Prof. Dr. Zorah Lähner

**Programme**

M. Sc. Computer Science

**Mode**

Optional

**Semester**

2.

**Learning goals: technical skills**

Students will carry out a practical task (project) in the context of visual computing, including test and documentation of the implemented software/system.

**Learning goals: soft skills**

Ability to properly present and defend design decisions, to prepare readable documentation of software; skills in constructively collaborating with others in small teams over a longer period of time; ability to classify ones own results into the state-of-the-art of the respective area.

**Contents**

This lab introduces methods and applications in the field of geometry processing. You will get a chance to study the methods in depth by implementing them and running experiments. At the end of the semester, you will present the method, give a short demonstration and hand in a report describing the method and experimental outcomes.

**Prerequisites**

none

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lab</td>
<td>8</td>
<td>4</td>
<td>60 T / 210 S</td>
<td>9</td>
</tr>
</tbody>
</table>

T = face-to-face teaching  
S = independent study

**Graded exams**

Oral presentation, written report

**Ungraded coursework (required for admission to the exam)**
MA-INF 2227  Lab 3D Animation

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>every semester</td>
</tr>
</tbody>
</table>

**Module coordinator**

Prof. Dr. Ina Prinz

**Lecturer(s)**

Prof. Dr. Ina Prinz

**Programme**

M. Sc. Computer Science

**Mode**

Optional

**Semester**

1-3.

**Learning goals: technical skills**

The students will carry out a practical task (project) in the context of 3D animation, containing modelling, preparing a screenplay, realizing an animation related to real physical laws, rendering and creating a video.

**Learning goals: soft skills**

Ability to properly present and defend design decisions, to prepare readable documentation of software; skills in constructively collaborating with others in small teams over a longer period of time; ability to classify one’s own results into the state-of-the-art of the resp. area.

**Contents**

Varying selected topics close to current research in the area of the history of computing and the mechanization of computing as well as deep understanding of mechanical and technical functions and its presentation in a representative 3D animation video, contains technical visualization and didactic skills.

**Prerequisites**

Recommended:

- BA-INF 108 Geschichte des maschinellen Rechnens I
- BA-INF 126 Geschichte des maschinellen Rechnens II

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lab</td>
<td>8</td>
<td>4</td>
<td>60 T / 210 S</td>
<td>9</td>
</tr>
</tbody>
</table>

T = face-to-face teaching

S = independent study

**Graded exams**

Oral presentation, written report, presentation of the video

**Ungraded coursework (required for admission to the exam)**
MA-INF 2307  Lab Vision

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>every semester</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Module coordinator</th>
<th>Lecturer(s)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Prof. Dr. Jürgen Gall</td>
<td>Prof. Dr. Jürgen Gall</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Programme</th>
<th>Mode</th>
<th>Semester</th>
</tr>
</thead>
<tbody>
<tr>
<td>M. Sc. Computer Science</td>
<td>Optional</td>
<td>2. or 3.</td>
</tr>
</tbody>
</table>

**Learning goals: technical skills**
The students will carry out a practical computer vision task (project).

**Learning goals: soft skills**
Ability to properly present and defend design decisions, to prepare readable documentation of software; skills in constructively collaborating with others in small teams over a longer period of time; ability to classify own results into the state-of-the-art of the resp. area

**Contents**
Computer Vision: research topics and applications

**Prerequisites**
Required:
MA-INF 2201 – Computer Vision
Good C++ or Python programming skills

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lab</td>
<td>8</td>
<td>4</td>
<td>60 T / 210 S</td>
<td>9</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

**Graded exams**
Oral presentation, written report

**Ungraded coursework (required for admission to the exam)**
MA-INF 2308  Lab Graphics

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>every semester</td>
</tr>
</tbody>
</table>

**Module coordinator**
Prof. Dr. Reinhard Klein

**Lecturer(s)**
Prof. Dr. Reinhard Klein

**Programme**
M. Sc. Computer Science

**Mode**
Optional

**Semester**
3.

**Learning goals: technical skills**
The students will carry out a practical task (project) in the context of geometry processing, rendering, scientific visualization or human computer interaction, including test and documentation of the implemented software/system.

**Learning goals: soft skills**
Ability to properly present and defend design decisions, to prepare readable documentation of software; skills in constructively collaborating with others in small teams over a longer period of time; ability to classify ones own results into the state-of-the-art of the resp. area.

**Contents**
Varying selected topics close to current research in the area of geometry processing, rendering, scientific visualization or human computer interaction.

**Prerequisites**
none

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lab</td>
<td>8</td>
<td>4</td>
<td>60 T / 210 S</td>
<td>9</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

**Graded exams**
Oral presentation, written report

**Ungraded coursework (required for admission to the exam)**
### MA-INF 2309  Lab Audio

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

**Module coordinator**
apl. Prof. Dr. Frank Kurth
Lecturer(s)
apl. Prof. Dr. Frank Kurth, Prof. Dr. Michael Clausen

**Programme**
M. Sc. Computer Science
Mode
Optional
Semester
3.

**Learning goals: technical skills**
The students will carry out a practical task (project) in the context of audio and music processing, including test and documentation of the implemented software/system.

**Learning goals: soft skills**
Ability to properly present and defend design decisions, to prepare readable documentation of software; skills in constructively collaborating with others in small teams over a longer period of time; ability to classify ones own results into the state-of-the-art of the resp. area.

**Contents**

**Prerequisites**
none

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lab</td>
<td>8</td>
<td>4</td>
<td>60 T / 210 S</td>
<td>9</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

**Graded exams**
Oral presentation, written report

**Ungraded coursework (required for admission to the exam)**
MA-INF 2310  Advanced Topics in Computer Graphics II

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

Module coordinator
Prof. Dr. Reinhard Klein

Lecturer(s)
Prof. Dr. Reinhard Klein

Programme
M. Sc. Computer Science

Mode
Optional

Semester
3.

Learning goals: technical skills
Analytical formulation of problems related to geometry processing:
- apply methods of geometry processing
- apply basic concepts of statistical shape analysis and shape spaces to real world applications
- Design and implement novel application software in this area

Learning goals: soft skills
Based on the knowledge and skills acquired students should be able to
- read and judge current scientific literature in the area of geometry processing and gain an overview of the current state of the art
- identify the major literature relevant for solving a given problem in geometry processing
- present, propose and communicate different solutions and work in a team to solve geometry processing problems
- discuss geometry processing problems with researchers from different application fields

Contents
This course will first introduce the mathematical and algorithmic tools required to represent, model, and process 3D geometric objects. The second part discusses the latest mathematical, algorithmic, and statistical tools required for the analysis and modeling of 3D shape variability, which can facilitate the creation of 3D models. Topics among others will be
- classical and discrete differential geometry of curves and surfaces
- mesh data structures and generation of meshes from point clouds
- Laplacian operator and optimization techniques with applications to denoising, smoothing, decimation, shape fitting, shape descriptors, geodesic distances
- parameterization and editing of surfaces
- point cloud registration
- correspondences
- shape spaces and statistical shape analysis

In addition, results from state-of-the-art research will be presented.

Prerequisites
none

Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td>4</td>
<td>60 T / 105 S</td>
<td>5.5</td>
<td></td>
</tr>
<tr>
<td>Exercises</td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
<td></td>
</tr>
</tbody>
</table>

Graded exams
Oral exam

Ungraded coursework (required for admission to the exam)
Successful exercise participation
Literature

- M. Botsch, L. Kobbelt, M. Pauly, P. Alliez, B. Levy, Polygon Mesh Processing, A K Peters, 2010
MA-INF 2312  Image Acquisition and Analysis in Neuroscience

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>180 h</td>
<td>6 CP</td>
<td>1 semester</td>
<td>at least every 2 years</td>
</tr>
</tbody>
</table>

**Module coordinator**
Prof. Dr. Thomas Schultz

**Lecturer(s)**
Prof. Dr. Thomas Schultz

**Programme**
M. Sc. Computer Science

**Mode**
Optional

**Semester**
1-4.

**Learning goals: technical skills**
Students will learn about image acquisition and analysis pipelines which are used in neuroscience. They will understand algorithms for image reconstruction, artifact removal, image registration and segmentation, as well as relevant statistical and machine learning techniques. A particular focus will be on data from Magnetic Resonance Imaging and on mathematical models for functional and diffusion MRI data.

**Learning goals: soft skills**
Productive work in small teams, self-dependent solution of practical problems in the area of biomedical image processing, presentation of solution strategies and implementations, self management, critical reflection of conclusions drawn from complex experimental data.

**Contents**
This course covers the full image formation and analysis pipeline that is typically used in biomedical studies, from image acquisition to image processing and statistical analysis.

**Prerequisites**
Recommended:
Mathematical background (calculus, linear algebra, statistics); imperative programming.

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td>3</td>
<td>45 T / 45 S</td>
<td>3</td>
<td></td>
</tr>
<tr>
<td>Exercises</td>
<td>1</td>
<td>15 T / 75 S</td>
<td>3</td>
<td></td>
</tr>
</tbody>
</table>

**T** = face-to-face teaching  
**S** = independent study

**Graded exams**
Oral exam

**Ungraded coursework (required for admission to the exam)**
Successful exercise participation

**Literature**
MA-INF 2313  Deep Learning for Visual Recognition

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>180 h</td>
<td>6 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

Module coordinator
Prof. Dr. Reinhard Klein

Lecturer(s)
Dr. Michael Weinmann

Programme
M. Sc. Computer Science

Mode
Optional

Semester
1-4.

Learning goals: technical skills
Students will be introduced to the theory of neural networks and study various applications in computer vision and other topics in AI.

Learning goals: soft skills
Productive work in small teams, development and realization of individual approaches and solutions, critical reflection of competing methods, discussion in groups.

Contents
Deep learning has taken over the machine learning community by storm, with success both in research and commercially. Deep learning is applicable over a range of fields such as computer vision, speech recognition, natural language processing, robotics, etc. This course will introduce the fundamentals of neural networks and then progress to state-of-the-art convolutional and recurrent neural networks as well as their use in applications for visual recognition. Students will get a chance to learn how to implement and train their own network for visual recognition tasks such as object recognition, image segmentation and caption generation.

Prerequisites
Recommended:
Students are recommended to have a basic knowledge in probability and statistics and linear algebra as well as proficiency in programming (python or Matlab or C++).

Course meetings

| Teaching format | Group size | h/week | Workload|h | CP | T = face-to-face teaching | S = independent study |
|----------------|------------|--------|---------|-----|-------------------------|-----------------------|
| Lecture        | 2          | 30 T / 45 S | 2.5   |     |                         |                       |
| Exercises      | 2          | 30 T / 75 S | 3.5   |     |                         |                       |

Graded exams

Oral exam

Ungraded coursework (required for admission to the exam)
Successful exercise participation

Literature
No required text. Supplemental readings will be provided in the lecture.
MA-INF 2315  Seminar Computational Photography

<table>
<thead>
<tr>
<th></th>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>120 h</td>
<td>4 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

**Module coordinator**
Prof. Dr. Matthias Hullin

**Lecturer(s)**
Prof. Dr. Matthias Hullin

**Programme**
M. Sc. Computer Science

**Mode**
Optional

**Semester**
2. or 3.

**Learning goals: technical skills**
Ability to understand new research results presented in original scientific papers.

**Learning goals: soft skills**
Ability to present and to critically discuss these results in the framework of the corresponding area.

**Contents**

**Prerequisites**
none

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Seminar</td>
<td>10</td>
<td>2</td>
<td>30 T / 90 S</td>
<td>4</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

**Graded exams**

Oral presentation, written report

**Ungraded coursework (required for admission to the exam)**
MA-INF 2316  Lab Digital Material Appearance

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

Module coordinator
Prof. Dr. Matthias Hullin

Lecturer(s)
Prof. Dr. Matthias Hullin

Programme
M. Sc. Computer Science

Mode
Optional

Semester
2. or 3.

Learning goals: technical skills
The students will carry out a practical task (project) in the context of the corresponding area, including test and documentation of the implemented software/system.

Learning goals: soft skills
Ability to properly present and defend design decisions, to prepare readable documentation of software; skills in constructively collaborating with others in small teams over a longer period of time; ability to classify ones own results into the state-of-the-art of the resp. area

Contents

Prerequisites
none

Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lab</td>
<td>8</td>
<td>4</td>
<td>60 T / 210 S</td>
<td>9</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

Graded exams
Oral presentation, written report

Ungraded coursework (required for admission to the exam)

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>180 h</td>
<td>6 CP</td>
<td>1 semester</td>
<td>at least every 2 years</td>
</tr>
</tbody>
</table>

Module coordinator
Prof. Dr. Florian Bernard

Lecturer(s)
Prof. Dr. Florian Bernard

Programme
M. Sc. Computer Science

Mode
Optional

Semester
2. or 3.

Learning goals: technical skills
- ability to implement basic numerical algorithms, understanding their strengths and shortcomings
- mathematical modelling of computational problems in visual computing and machine learning
- gain an intuition which algorithm is best applied for which problem in visual computing and machine learning, so that practical problems in these areas can be solved

Learning goals: soft skills
- problem solving skills: ability to identify and utilise analogies between new problems and previously seen ones
- analytical and abstract thinking: develop a general intuition of computational problems, being able to adopt different perspectives of particular concepts

Contents
This module focuses on numerical methods that frequently occur in the fields visual computing (VC) and machine learning (ML). In addition to algorithms, this module will also cover modelling aspects that are relevant for solving practical problems in VC and ML. The contents include:

- Error analysis and conditioning of problems
- Linear systems (solvability, algorithms, stability, regularisation), and applications and modelling in VC and ML (e.g. linear regression, image alignment, deconvolution)
- Spectral methods (eigenvalue decomposition, singular value decomposition, respective algorithms), and their applications and modelling in VC and ML (e.g. clustering, Procrustes analysis, point-cloud alignment, principal components analysis)
- Numerical optimisation (gradient-based methods, second-order methods, large-scale optimisation) and applications and modelling in VC and ML.

Prerequisites
Required:
No formal prerequisites.

Recommended:
Participants are expected to have a high level of mathematical maturity (in particular, a good working knowledge of linear algebra and calculus is essential). A basic understanding of mathematical optimisation is advantageous.

Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td>2</td>
<td>30 T / 45 S</td>
<td>2.5</td>
<td></td>
</tr>
<tr>
<td>Exercises</td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
<td></td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

Graded exams
Written exam

Ungraded coursework (required for admission to the exam)
Successful exercise participation
## 3 Information and Communication Management

<table>
<thead>
<tr>
<th>Course Code</th>
<th>Type</th>
<th>Credits</th>
<th>Title</th>
</tr>
</thead>
<tbody>
<tr>
<td>MA-INF 3108</td>
<td>L2E2</td>
<td>6 CP</td>
<td>Secure Software Engineering</td>
</tr>
<tr>
<td>MA-INF 3109</td>
<td>L2E2</td>
<td>6 CP</td>
<td>Quantum Algorithms: Introduction and Data Fusion Examples</td>
</tr>
<tr>
<td>MA-INF 3140</td>
<td>L2E2</td>
<td>6 CP</td>
<td>Advanced Computer Forensics</td>
</tr>
<tr>
<td>MA-INF 3202</td>
<td>L2E2</td>
<td>6 CP</td>
<td>Mobile Communication</td>
</tr>
<tr>
<td>MA-INF 3209</td>
<td>Sem2</td>
<td>4 CP</td>
<td>Seminar Selected Topics in Communication Management</td>
</tr>
<tr>
<td>MA-INF 3216</td>
<td>Sem2</td>
<td>4 CP</td>
<td>Seminar Sensor Data Fusion</td>
</tr>
<tr>
<td>MA-INF 3229</td>
<td>Lab4</td>
<td>9 CP</td>
<td>Lab IT-Security</td>
</tr>
<tr>
<td>MA-INF 3233</td>
<td>L2E2</td>
<td>6 CP</td>
<td>Advanced Sensor Data Fusion in Distributed Systems</td>
</tr>
<tr>
<td>MA-INF 3236</td>
<td>L2E2</td>
<td>6 CP</td>
<td>IT Security</td>
</tr>
<tr>
<td>MA-INF 3237</td>
<td>L2E2</td>
<td>6 CP</td>
<td>Array Signal and Multi-channel Processing</td>
</tr>
<tr>
<td>MA-INF 3238</td>
<td>L2E2</td>
<td>6 CP</td>
<td>Side Channel Attacks</td>
</tr>
<tr>
<td>MA-INF 3239</td>
<td>L2E2</td>
<td>6 CP</td>
<td>Malware Analysis</td>
</tr>
<tr>
<td>MA-INF 3241</td>
<td>L3E1</td>
<td>6 CP</td>
<td>Practical Challenges in Human Factors of Security and Privacy</td>
</tr>
<tr>
<td>MA-INF 3242</td>
<td>L2E2</td>
<td>6 CP</td>
<td>Security of Distributed and Resource-constrained Systems</td>
</tr>
<tr>
<td>MA-INF 3304</td>
<td>Lab4</td>
<td>9 CP</td>
<td>Lab Communication and Communicating Devices</td>
</tr>
<tr>
<td>MA-INF 3305</td>
<td>Lab4</td>
<td>9 CP</td>
<td>Lab Information Systems</td>
</tr>
<tr>
<td>MA-INF 3309</td>
<td>Lab4</td>
<td>9 CP</td>
<td>Lab Malware Analysis</td>
</tr>
<tr>
<td>MA-INF 3310</td>
<td>L2E2</td>
<td>6 CP</td>
<td>Introduction to Sensor Data Fusion - Methods and Applications</td>
</tr>
<tr>
<td>MA-INF 3312</td>
<td>Lab4</td>
<td>9 CP</td>
<td>Lab Sensor Data Fusion</td>
</tr>
<tr>
<td>MA-INF 3317</td>
<td>Sem2</td>
<td>4 CP</td>
<td>Seminar Selected Topics in IT Security</td>
</tr>
<tr>
<td>MA-INF 3319</td>
<td>Lab4</td>
<td>9 CP</td>
<td>Lab Usable Security and Privacy</td>
</tr>
<tr>
<td>MA-INF 3320</td>
<td>Lab4</td>
<td>9 CP</td>
<td>Lab Security in Distributed Systems</td>
</tr>
<tr>
<td>MA-INF 3321</td>
<td>Sem2</td>
<td>4 CP</td>
<td>Seminar Usable Security and Privacy</td>
</tr>
<tr>
<td>MA-INF 3322</td>
<td>L2E2</td>
<td>6 CP</td>
<td>Applied Binary Exploitation</td>
</tr>
<tr>
<td>MA-INF 3323</td>
<td>Lab4</td>
<td>9 CP</td>
<td>Lab Fuzzing Bootcamp</td>
</tr>
<tr>
<td>MA-INF 3324</td>
<td>Lab4</td>
<td>9 CP</td>
<td>Lab Design of Usable Security Mechanisms</td>
</tr>
</tbody>
</table>
MA-INF 3108  Secure Software Engineering

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>180 h</td>
<td>6 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Module coordinator</th>
<th>Lecturer(s)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Dr. Christian Tiefenau</td>
<td>Dr. Christian Tiefenau, Mischa Meier</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Programme</th>
<th>Mode</th>
<th>Semester</th>
</tr>
</thead>
<tbody>
<tr>
<td>M. Sc. Computer Science</td>
<td>Optional</td>
<td>2. or 3.</td>
</tr>
</tbody>
</table>

**Learning goals: technical skills**
The students are introduced to the security-relevant aspects of a software-engineering lifecycle. Therefore, the main ideas of including security throughout the development process will be presented and explained by examples.

By showing common vulnerabilities throughout this course, the students will get an understanding of common vulnerabilities and attacks and how to prevent them.

**Learning goals: soft skills**
In groups, the students will conduct practical exercises to strengthen the understanding of vulnerabilities and attack vectors. Through this, the abilities teamwork, time management, organization and critical discussion of their own and others’ results are strengthened.

**Contents**
- Threat modeling
- Risk analysis
- Architectural security
- Secure coding
- Applied Cryptography
- Secure configuration and deployment
- Updates and maintenance

**Prerequisites**
Recommended:
Fundamental knowledge in software-engineering and IT-security concepts.

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td>2</td>
<td>30 T / 45 S</td>
<td>2.5</td>
<td></td>
</tr>
<tr>
<td>Exercises</td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
<td></td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

**Graded exams**
Written exam

**Ungraded coursework (required for admission to the exam)**
Successful exercise participation

**Literature**
Software Security: Building Security In by Gary McGraw
MA-INF 3109  Quantum Algorithms: Introduction and Data Fusion
Examples

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>180 h</td>
<td>6 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

**Module coordinator**
Prof. Dr. Wolfgang Koch

**Lecturer(s)**
Prof. Dr. Wolfgang Koch, Dr. Felix Govaers, Dr. Martin Ulmke

**Programme**
M. Sc. Computer Science

**Mode**
Optional

**Semester**
2. or 3.

**Learning goals: technical skills**
Quantum algorithms for data fusion may become game changers as soon as quantum processing kernels embedded in hybrid processing architectures with classical processors will exist. While emerging quantum technologies directly apply quantum physics, quantum algorithms do not exploit quantum physical phenomena as such, but rather use the sophisticated framework of quantum physics to deal with “uncertainty”. Although the link between mathematical statistics and quantum physics has long been known, the potential of physics-inspired algorithms for data fusion has just begun to be realized. While the implementation of quantum algorithms is to be considered on classical as well as on quantum computers, the latter are anticipated as well-adapted “analog computers” for unprecedentedly fast solving data fusion and resources management problems. While the development of quantum computers cannot be taken for granted, their potential is nonetheless real and has to be considered by the international information fusion community.

**Learning goals: soft skills**
- Problem solving
- Adaptability
- Critical thinking

**Contents**
- Introduction with Examples
- Short introduction to quantum mechanics
- Introduction to quantum computing
- Quantum computing hardware
- Quantum inspired tracking
- Particle filtering and fermionic target tracking
- The data association problem
- Track extraction and sensor management
- Quantum computing for multi target tracking data association
- Quantum computing for resources management
- Quantum many particle systems and boson sampling
- Path Integrals

**Prerequisites**
Recommended:
One of the following:
- BA-INF 137 – Einführung in die Sensordatenfusion
- MA-INF 3310 – Introduction to Sensor Data Fusion - Methods and Applications

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td>2</td>
<td>30 T / 45 S</td>
<td>2.5</td>
<td></td>
</tr>
<tr>
<td>Exercises</td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
<td></td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

**Graded exams**
Oral exam

**Ungraded coursework (required for admission to the exam)**
Successful exercise participation
MA-INF 3140  Advanced Computer Forensics

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>180 h</td>
<td>6 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

Module coordinator: Dr. Christian Tiefenau
Lecturer(s): Dr. Christian Tiefenau

Programme: M. Sc. Computer Science
Mode: Optional
Semester: 1., 2. or 3.

Learning goals: technical skills
The course covers advanced research topics in computer forensics and secure software engineering.

Learning goals: soft skills

Contents
Theoretical and practical aspects of computer forensics and secure software engineering are covered.

Prerequisites
none

Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload</th>
<th>CP</th>
<th>T = face-to-face teaching</th>
<th>S = independent study</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td></td>
<td>2</td>
<td>30 T / 45 S</td>
<td>2.5</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Exercises</td>
<td></td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

Graded exams
Written exam

Ungraded coursework (required for admission to the exam)
Successful exercise participation
MA-INF 3202  Mobile Communication

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>180 h</td>
<td>6 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

**Module coordinator**
Prof. Dr. Peter Martini

**Lecturer(s)**
Prof. Dr. Peter Martini, Dr. Matthias Frank

**Programme**
M. Sc. Computer Science

**Mode**
Optional

**Semester**
2. or 3.

**Learning goals: technical skills**
Knowledge about key concepts of mobile communication including mobility management (both technology independent and technology dependent), knowledge about wireless technologies and their interaction with other protocol layers and/or other network technologies, ability to evaluate and assess scenarios with communication of mobile devices. In-depth understanding of communication paradigms of wireless/mobile systems and network elements, productive work in small groups, strengthening skills on presentation and discussion of solutions to current challenges.

**Learning goals: soft skills**
Theoretical exercises to support in-depth understanding of lecture topics and to stimulate discussions, practical exercises in teamwork to support time management, targeted organisation of practical work and critical discussion of own and others’ results.

**Contents**
Mobility Management in the Internet, Wireless Communication Basics, Wireless Networking Technologies, Cellular/Mobile Communication Networks (voice and data communication), Ad-hoc and Sensor Networks.

**Prerequisites**
Recommended:
Bachelor level knowledge of basics of communication systems (e.g. BA-INF 101 'Kommunikation in Verteilten Systemen' (German Bachelor Programme Informatik, English lecture slides available)

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload [h]</th>
<th>CP</th>
<th>T = face-to-face teaching</th>
<th>S = independent study</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td></td>
<td></td>
<td>30 T / 45 S</td>
<td>2.5</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Exercises</td>
<td></td>
<td></td>
<td>30 T / 75 S</td>
<td>3.5</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

**Graded exams**
Written exam

**Ungraded coursework (required for admission to the exam)**
The completion of regularly provided exercise sheets. The work can be done in groups of up to three students. A total of 50% of the points must be achieved. For 70% of the exercise sheets, 20% of the points must be achieved for each sheet.

**Literature**
- Jochen Schiller: Mobile Communications, Addison-Wesley, 2003
- Further up-to-date literature will be announced in due course before the beginning of the lecture.
# MA-INF 3209  Seminar Selected Topics in Communication Management

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>120 h</td>
<td>4 CP</td>
<td>1 semester</td>
<td>at least every year</td>
</tr>
</tbody>
</table>

**Module coordinator**

Prof. Dr. Peter Martini

**Lecturer(s)**

Prof. Dr. Peter Martini, Prof. Dr. Michael Meier

**Programme**

M. Sc. Computer Science

**Mode**

Optional

**Semester**

2. or 3.

---

## Learning goals: technical skills

Ability to understand new research results presented in original scientific papers.

## Learning goals: soft skills

Ability to present and to critically discuss these results in the framework of the corresponding area.

## Contents

Current conference and journal papers, current standardization drafts

## Prerequisites

**Required:**

Successful completion of at least one of the following lectures: Principles of Distributed Systems (MA-INF3105), Network Security (MA-INF3201), Mobile Communication (MA-INF3202), IT Security (MA-INF3236)

## Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Seminar</td>
<td>10</td>
<td>2</td>
<td>30 T / 90 S</td>
<td>4</td>
</tr>
</tbody>
</table>

T = face-to-face teaching

S = independent study

## Graded exams

Oral presentation, written report

**Ungraded coursework (required for admission to the exam)**

## Literature

The relevant literature will be announced towards the end of the previous semester
MA-INF 3216  Seminar Sensor Data Fusion

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>120 h</td>
<td>4 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

**Module coordinator**

P.D. Dr. Wolfgang Koch

**Lecturer(s)**

P.D. Dr. Wolfgang Koch, Dr. Felix Govaers

**Programme**

M. Sc. Computer Science

**Mode**

Optional

**Semester**

2.

**Learning goals: technical skills**

Ability to understand new research results presented in original scientific papers.

**Learning goals: soft skills**

Ability to present and to critically discuss these results in the framework of the corresponding area.

**Contents**

Current conference and journal papers

**Prerequisites**

none

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Seminar</td>
<td>10</td>
<td>2</td>
<td>30 T / 90 S</td>
<td>4</td>
</tr>
</tbody>
</table>

T = face-to-face teaching

S = independent study

**Graded exams**

Oral presentation, written report

**Ungraded coursework (required for admission to the exam)**

**Literature**

The relevant literature will be announced at the beginning of the seminar.
MA-INF 3229  Lab IT-Security

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>every semester</td>
</tr>
</tbody>
</table>

**Module coordinator**
Prof. Dr. Michael Meier

**Lecturer(s)**
Prof. Dr. Michael Meier

**Programme**
M. Sc. Computer Science

**Mode**
Optional

**Semester**
2. or 3.

**Learning goals: technical skills**
The students will carry out a practical task (project) in the context of IT Security, including test and documentation of the implemented software/system.

**Learning goals: soft skills**
Ability to properly present and defend design decisions, to prepare readable documentation of software; skills in constructively collaborating with others in small teams over a longer period of time; ability to classify ones own results into the state-of-the-art of the resp. area

**Contents**

**Prerequisites**
none

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lab</td>
<td>8</td>
<td>4</td>
<td>60 T / 210 S</td>
<td>9</td>
</tr>
</tbody>
</table>

T = face-to-face teaching  
S = independent study

**Graded exams**
Oral presentation, written report

**Ungraded coursework (required for admission to the exam)**
### MA-INF 3233 Advanced Sensor Data Fusion in Distributed Systems

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>180 h</td>
<td>6 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

**Module coordinator**
PD Dr. Wolfgang Koch

**Lecturer(s)**
Dr. Felix Govaers

**Programme**
M. Sc. Computer Science

**Frequency**
every year

**Mode**
Optional

**Semester**
2.

**Learning goals: technical skills**
For challenging state estimation tasks, algorithms which enhance the situational awareness by fusing sensor information are inevitable. Nowadays it has become very popular to improve the performance of systems by linking multiple sensors. This implies some challenges to the sensor data fusion methodologies such as sensor registration, communication delays, and correlations of estimation errors. In particular, if the communication links have limited bandwidth, data reduction techniques have to be applied at the sensor sites, that is local tracks have to be computed. Once received at a fusion center (FC), the tracks then are fused to reconstruct a global estimate. In this lecture, methodologies to achieve a distributed state estimation are considered. Among these are tracklet fusion, the Bar-Shalom-Campo formula, the Federated Kalman Filter, naive fusion, the distributed Kalman filter and the least squares estimate.

**Learning goals: soft skills**
Mathematical derivation of algorithms, application of mathematical results on estimation theory.

**Contents**
tracklet fusion, the Bar-Shalom-Campo formula, the Federated Kalman Filter, naive fusion, the distributed Kalman filter and the least squares estimate, Accumulated State Densities, Decorrelated fusion, product representation

**Prerequisites**
Recommended:
At least 1 of the following:
BA-INF 137 – Einführung in die Sensordatenfusion
MA-INF 3310 – Introduction to Sensor Data Fusion - Methods and Applications

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td>2</td>
<td>30 T</td>
<td>45 S</td>
<td>2.5</td>
</tr>
<tr>
<td>Exercises</td>
<td>2</td>
<td>30 T</td>
<td>75 S</td>
<td>3.5</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

**Graded exams**
Oral exam

**Ungraded coursework (required for admission to the exam)**
50% of the maximum achievable points in the practical programming exercises are required. The delivery of the programmed solution is done individually or in group work of up to three students. A total of 10 points will be awarded, 50% of which will have been achieved if the Distributed Kalman filter has been programmed in an executable and consistent manner.

**Forms of media**
Power Point

**Literature**
MA-INF 3236  IT Security

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>180 h</td>
<td>6 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

Module coordinator
Prof. Dr. Michael Meier

Lecturer(s)
Prof. Dr. Michael Meier

Programme
M. Sc. Computer Science

Mode
Optional

Semester
1. or 2.

Learning goals: technical skills
Students are introduced to a variety of active research fields in IT security. Students learn about the motivation, challenges and objectives in these fields. Additionally, they get to know selected fundamental knowledge and methods helping them to deepen their knowledge in their upcoming studies.

Learning goals: soft skills
working in small groups on exercises, critical discussion of own and others’ results, time management, transferring theoretical knowledge to practical scenarios

Contents
The contents vary but usually include
- Privacy
- Cryptographic Protocols
- Network Security
- Supply Chain Attacks
- Management of Identity Data
- Low-level software analysis
- Software testing
- Side Channel Attacks
- Anomaly Detection
- Human Factor in Security

Prerequisites
Required:
Fundamental knowledge in the following areas: operating systems, networks, security

Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td></td>
<td>2</td>
<td>30 T / 45 S</td>
<td>2.5</td>
</tr>
<tr>
<td>Exercises</td>
<td></td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

Graded exams
Written exam

Ungraded coursework (required for admission to the exam)
Successful exercise participation
MA-INF 3237  Array Signal and Multi-channel Processing

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>180 h</td>
<td>6 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

Module coordinator
Prof. Dr. Wolfgang Koch

Lecturer(s)
Dr. Marc Oispuu

Programme
M. Sc. Computer Science

Mode
Optional

Semester
2. or 3.

Learning goals: technical skills
Localization of multiple sources using passive sensors is a fundamental task encountered in various fields like wireless communication, radar, sonar, and seismology. In this lecture, a unified framework for electromagnetic and acoustic signals and signal processing techniques are presented. Furthermore, the sensor calibration, direction finding, and bearings-only localization problem are considered. Special applications are emphasized, like small airborne arrays for unmanned aerial vehicles (UAVs).

Learning goals: soft skills
Mathematical derivation of algorithms, applications of mathematical results on estimation theory

Contents
Estimation theory, Sensor model, Cramér-Rao analysis, conventional beamforming, Multiple Signal Classification (MUSIC), sensor calibration, Bearings-only localization, Direct Position Determination (DPD), Applications

Prerequisites
Recommended:

Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td>2</td>
<td>30 T / 45 S</td>
<td>2.5</td>
<td></td>
</tr>
<tr>
<td>Exercises</td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
<td></td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

Graded exams
Oral Exam

Ungraded coursework (required for admission to the exam)
50% of the maximum achievable points in the practical programming exercises are required. The delivery of the programmed solution is done individually or in group work of up to three students. A total of 10 points will be awarded, 50% of which will have been achieved if the basic signal processing algorithms for array sensors have been implemented.

Forms of media
Power Point

Literature
MA-INF 3238  Side Channel Attacks

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>180 h</td>
<td>6 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

**Module coordinator**
Dr. Felix Boes

**Lecturer(s)**
Dr. Felix Boes

**Programme**
M. Sc. Computer Science

**Mode**
Optional

**Semester**
2. or 3.

**Learning goals: technical skills**
- Students are introduced to theoretical and practical side channel effects of modern hardware.
- Students learn techniques to utilize these effects to circumvent security mechanisms.
- This includes covert channels as well as side channel attacks and microarchitectural attacks on modern CPUs.

**Learning goals: soft skills**
Theoretical exercises to support in-depth understanding of lecture topics and to stimulate discussions, practical exercises in teamwork to support time management, targeted organization of practical work and critical discussion of own and others’ results.

**Contents**
- Theoretical foundations of side channel effects and attacks as well as
  - covert channels,
  - differential power analysis,
  - padding oracle,
  - RSA timing attacks,
  - cache based side channel effects,
  - microarchitectural attacks (Spectre)

**Prerequisites**
Recommended:
Fundamental knowledge about IT Security, operating systems and statistics is advantageous but not mandatory.

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td>2</td>
<td>30 T / 45 S</td>
<td>2.5</td>
<td></td>
</tr>
<tr>
<td>Exercises</td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
<td></td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

**Graded exams**
Written Exam

**Ungraded coursework (required for admission to the exam)**
Successful exercise participation
MA-INF 3239  Malware Analysis

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>180 h</td>
<td>6 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

Module coordinator
Prof. Dr. Peter Martini

Lecturer(s)
Prof. Dr. Elmar Padilla

Programme
M. Sc. Computer Science

Mode
Optional

Semester
2. or 3.

Learning goals: technical skills
The students should be able to analyze the functional scope of a binary file independently and to describe its damage potential. In addition, the students should be able to carry out detailed analyzes of given aspects and to partially automate these with the help of scripts.

Learning goals: soft skills
Presentation of solutions and methods, critical discussion of applied methods and techniques.

Contents
In the course, the skills acquired so far in binary analysis will first be deepened and adapted to the peculiarities of malware analysis. Different malware samples are used to explain the techniques used by malware authors. These priorities include:

- Characteristics of malware
- Persistence
- Network communication
- Encryption
- Dynamic malware analysis
- Debugging
- Behavioral obfuscation
- Virtual analysis environments
- Static malware analysis
- Control flow obfuscation
- Automation of common analysis steps
- Reconstruction of binary algorithms

The event begins with several lectures that provide the basics for the students to work independently later. In the course of this, the students will work on practical topics from the field of malware analysis during the semester. Since these subject areas can turn out to be very specific, it is necessary to be willing to deal with the subject outside of the lecture and exercise times.

Prerequisites
Required: none

Recommended:
Basic knowledge of operating systems (kernel, threads, virtual memory), network communication (protocols, architectures), binary analysis (assembler, endianness, semantic gap, coding), software development (programming, semantics, scripting in Python)

Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td></td>
<td>2</td>
<td>30 T / 45 S</td>
<td>2.5</td>
</tr>
<tr>
<td>Exercises</td>
<td></td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

Graded exams

Oral exam

Ungraded coursework (required for admission to the exam)
The completion of regularly provided exercise sheets. The work can be done in groups of up to two students. A total of 50% of the points must be achieved.
<table>
<thead>
<tr>
<th>Literature</th>
</tr>
</thead>
<tbody>
<tr>
<td>The relevant literature will be announced at the beginning of the lecture</td>
</tr>
</tbody>
</table>
MA-INF 3241  Practical Challenges in Human Factors of Security and Privacy

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>180 h</td>
<td>6 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

**Module coordinator**  
Prof. Dr. Matthew Smith

**Lecturer(s)**  
Prof. Dr. Matthew Smith

**Programme**  
M. Sc. Computer Science

**Mode**  
Optional

**Semester**  
2.

**Learning goals: technical skills**
After completing the unit students will be able to conduct related work searches to get a deep understanding into the state of the art. They will be able to design, run and evaluate scientific studies in this area.

**Learning goals: soft skills**

**Contents**
In this course we will learn about and develop solutions for a specific challenge concerning human factors in security and privacy.

**Prerequisites**
none

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td>1</td>
<td>15 T / 45 S</td>
<td>2</td>
<td></td>
</tr>
<tr>
<td>Exercises</td>
<td>3</td>
<td>45 T / 75 S</td>
<td>4</td>
<td></td>
</tr>
</tbody>
</table>

T = face-to-face teaching  
S = independent study

**Graded exams**
Project work

**Ungraded coursework (required for admission to the exam)**
Successful exercise participation
MA-INF 3242  Security of Distributed and Resource-constrained Systems

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>180 h</td>
<td>6 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

**Module coordinator**
Prof. Dr. Michael Meier

**Lecturer(s)**
Dr. Thorsten Aurisch

**Programme**
M. Sc. Computer Science

**Mode**
Optional

**Semester**
2.

**Learning goals: technical skills**
Ability to understand and analyse theoretical and practical cyber security challenges of distributed and resource-constrained systems, as well as the ability to select and apply appropriate solutions.

**Learning goals: soft skills**

**Contents**
- Group communication with IP multicast
- Group key management
- Broadcast encryption
- Public key infrastructure
- Web of trust
- Multicast infrastructure protection
- Distributed security mechanisms
- Cyber resilience in groups
- Security in tactical radio networks
- Security for IoT

**Prerequisites**
none

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td>2</td>
<td>30 T / 45 S</td>
<td>2.5</td>
<td></td>
</tr>
<tr>
<td>Exercises</td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
<td></td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

**Graded exams**
Written exam

**Ungraded coursework (required for admission to the exam)**
Successful exercise participation
MA-INF 3304  Lab Communication and Communicating Devices

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>every semester</td>
</tr>
</tbody>
</table>

**Module coordinator**  
Prof. Dr. Peter Martini

**Lecturer(s)**  
Prof. Dr. Peter Martini, Prof. Dr. Michael Meier

**Programme**  
M. Sc. Computer Science

**Mode**  
Optional

**Semester**  
2. or 3.

### Learning goals: technical skills

The students will carry out a practical task (project) in the context of communication systems, including test and documentation of the implemented software/system.

### Learning goals: soft skills

Work in small teams and cooperate with other teams in a group; ability to make design decisions in a practical task; present and discuss (interim and final) results in the team/group and to other students; prepare written documentation of the work carried out.

### Contents

Selected topics close to current research in the area of communication systems, network security, mobile communication and communicating devices.

### Prerequisites

**Required:**  
Successful completion of at least one of the following lectures: Principles of Distributed Systems (MA-INF3105), Network Security (MA-INF3201), Mobile Communication (MA-INF3202), IT Security (MA-INF3236)

### Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lab</td>
<td>8</td>
<td>4</td>
<td>60 T / 210 S</td>
<td>9</td>
</tr>
</tbody>
</table>

T = face-to-face teaching  
S = independent study

### Graded exams

Oral presentation, written report

### Ungraded coursework (required for admission to the exam)

### Literature

The relevant literature will be announced towards the end of the previous semester.
MA-INF 3305  Lab Information Systems

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>at least every year</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Module coordinator</th>
<th>Lecturer(s)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Dr. Thomas Bode</td>
<td>Dr. Thomas Bode</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Programme</th>
<th>Mode</th>
<th>Semester</th>
</tr>
</thead>
<tbody>
<tr>
<td>M. Sc. Computer Science</td>
<td>Optional</td>
<td>2. or 3.</td>
</tr>
</tbody>
</table>

**Learning goals: technical skills**
The students will carry out a practical task (project) in the context of information systems, including test and documentation of the implemented software/system.

**Learning goals: soft skills**
Ability to properly present and defend design decisions, to prepare readable documentation of software; skills in constructively collaborating with others in small teams over a longer period of time; ability to classify ones own results into the state-of-the-art of the resp. area

**Contents**
Varying selected topics close to current research in the area of database- and information systems.

**Prerequisites**
none

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lab</td>
<td>8</td>
<td>4</td>
<td>60 T / 210 S</td>
<td>9</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

**Graded exams**
Oral presentation, written report

**Ungraded coursework (required for admission to the exam)**

**Literature**
The relevant literature will be announced towards the end of the previous semester.
**MA-INF 3309  Lab Malware Analysis**

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>every semester</td>
</tr>
</tbody>
</table>

**Module coordinator**

Prof. Dr. Peter Martini

**Lecturer(s)**

Prof. Dr. Peter Martini, Prof. Dr. Michael Meier

**Programme**

M. Sc. Computer Science

**Mode**

Optional

**Semester**

3.

**Learning goals: technical skills**

The students will carry out a practical task (project) in the context of communication systems with a specific topic focus on Malware Analysis and Computer/Network Security, including test and documentation of the implemented software/system.

**Learning goals: soft skills**

Work in small teams and cooperate with other teams in a group; ability to make design decisions in a practical task; present and discuss (interim and final) results in the team/group and to other students; prepare written documentation of the work carried out.

**Contents**

Selected topics close to current research in the area of communication systems, malware analysis, computer and network security.

**Prerequisites**

**Required:**

Successful completion of at least one of the following lectures: Principles of Distributed Systems (MA-INF3105), Network Security (MA-INF3201), Mobile Communication (MA-INF3202), IT Security (MA-INF3236)

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lab</td>
<td>8</td>
<td>4</td>
<td>60 T / 210 S</td>
<td>9</td>
</tr>
</tbody>
</table>

T = face-to-face teaching  
S = independent study

**Graded exams**

Oral presentation, written report

**Ungraded coursework (required for admission to the exam)**
Introduction to Sensor Data Fusion - Methods and Applications

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>180 h</td>
<td>6 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

Module coordinator: Prof. Dr. Wolfgang Koch  
Lecturer(s): Prof. Dr. Wolfgang Koch

Programme: M. Sc. Computer Science  
Mode: Optional  

Learning goals: technical skills
All participants shall get known to the basic theory of sensor data fusion. The lecture starts with preliminaries on how to handle uncertain data and knowledge within analytical calculus. Then, the fundamental and well-known Kalman filter is derived. Based on this tracking scheme, further approaches to a wide spectrum of applications will be shown. All algorithms will be motivated by examples from ongoing research projects, industrial cooperations, and impressions of current demonstration hardware. Because of inherent practical issues, every sensor measures certain properties up to an error. This lecture shows how to model and overcome this error by an application of theoretical tools such as Bayes’ rule and further derivations. Moreover, solutions to possible false-alarms, miss-detections, maneuvering phases, and much more will be presented.

Learning goals: soft skills
Mathematical derivation of algorithms, application of mathematical results on estimation theory.

Contents
Gaussian probability density functions, Kalman filter, Unscented Kalman Filter, Extended Kalman Filter, Particle Filter, Multi-Hypothesis-Trackier, Extended Target Tracking, Road Tracking, Interacting Multiple Model Filter, Retrodiction, Smoothing, Maneuver Modeling

Prerequisites
none

Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td>2</td>
<td>30 T / 45 S</td>
<td>2.5</td>
<td></td>
</tr>
<tr>
<td>Exercises</td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
<td></td>
</tr>
</tbody>
</table>

Graded exams
Written exam

Ungraded coursework (required for admission to the exam)
Successful exercise participation

Literature
MA-INF 3312  Lab Sensor Data Fusion

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

**Module coordinator**
Prof. Dr. Wolfgang Koch

**Lecturer(s)**
Prof. Dr. Wolfgang Koch

**Programme**
M. Sc. Computer Science

**Mode**
Optional

**Semester**
3.

**Learning goals: technical skills**
The students will work together on a data fusion project using various sensor hardware. Latest algorithms for fusing information from several nodes will be implemented.

**Learning goals: soft skills**
The students shall work together in a team. Everyone is responsible for a specific part in the context of a main goal. Results will be exchanged and integrated via software interfaces.

**Contents**
Varying selected topics on sensor data fusion.

**Prerequisites**
one

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lab</td>
<td>8</td>
<td>4</td>
<td>60 T / 210 S</td>
<td>9</td>
</tr>
</tbody>
</table>

**Graded exams**
Oral presentation, written report

**Ungraded coursework (required for admission to the exam)**

**Literature**
The relevant literature will be announced at the beginning of the lab.
# MA-INF 3317 Seminar Selected Topics in IT Security

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>120 h</td>
<td>4 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Module coordinator</th>
<th>Lecturer(s)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Prof. Dr. Michael Meier</td>
<td>Prof. Dr. Michael Meier, Prof. Dr. Peter Martini</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Programme</th>
<th>Mode</th>
<th>Semester</th>
</tr>
</thead>
<tbody>
<tr>
<td>M. Sc. Computer Science</td>
<td>Optional</td>
<td>2.</td>
</tr>
</tbody>
</table>

## Learning goals: technical skills

Ability to understand new research results presented in original scientific papers.

## Learning goals: soft skills

Ability to present and to critically discuss these results in the framework of the corresponding area.

## Contents

Current conference and journal papers

## Prerequisites

none

## Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
<th>T = face-to-face teaching</th>
<th>S = independent study</th>
</tr>
</thead>
<tbody>
<tr>
<td>Seminar</td>
<td>10</td>
<td>2</td>
<td>30 T / 90 S</td>
<td>4</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

## Graded exams

Oral presentation, written report

## Ungraded coursework (required for admission to the exam)
### MA-INF 3319 Lab Usable Security and Privacy

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

**Module coordinator**  
Prof. Dr. Matthew Smith  
**Lecturer(s)**  
Prof. Dr. Matthew Smith

**Programme**  
M. Sc. Computer Science  
**Mode**  
Optional  
**Semester**  
2.

**Learning goals: technical skills**  
The students will carry out a practical task (project) in the context of usable security and privacy, including user studies.

**Learning goals: soft skills**  
Ability to create and defend a scientific user study

**Contents**  
Students have a great degree of freedom to choose their own topics within the context of human aspects of security and privacy.

**Prerequisites**  
**Required:**  
Knowledge on how to run and evaluate user studies is required, for example as it is taught in BA-INF 145 - Usable Security and Privacy.

**Course meetings**  

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lab</td>
<td></td>
<td>4</td>
<td>60 T / 210 S</td>
<td>9</td>
</tr>
</tbody>
</table>

T = face-to-face teaching  
S = independent study

**Graded exams**  
Oral presentation, written report

**Ungraded coursework (required for admission to the exam)**
MA-INF 3320  Lab Security in Distributed Systems

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

**Module coordinator**
Prof. Dr. Matthew Smith

**Lecturer(s)**
Prof. Dr. Matthew Smith

**Programme**
M. Sc. Computer Science

**Mode**
Optional

**Semester**
2.

**Learning goals: technical skills**
The students will carry out a practical task (project) in the context of distributed security, including documentation of the implemented software/system.
Strong programming skills required.

**Learning goals: soft skills**
Ability to properly present and defend design decisions, to prepare readable documentation of software; skills in constructively collaborating with others in small teams over a longer period of time; ability to classify ones own results into the state-of-the-art of the resp. area

**Contents**
Security in distributed systems, including amongst others:
- Secure Messaging
- App Security
- SSL/HTTPS
- API Security
- Machine Learning for Security
- Passwords
- Intrusion Detection Systems
- Anomaly Detection
- Security Visualisation

**Prerequisites**
none

**Course meetings**

<table>
<thead>
<tr>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>8</td>
<td>4</td>
<td>60 T / 210 S</td>
<td>9</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

**Graded exams**
Oral presentation, written report

**Ungraded coursework (required for admission to the exam)**
# Seminar Usable Security and Privacy

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>120 h</td>
<td>4 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

## Module coordinator
- Prof. Dr. Matthew Smith

## Lecturer(s)
- Prof. Dr. Matthew Smith

## Programme
- M. Sc. Computer Science

## Mode
- Optional

## Semester
- 2.

## Learning goals: technical skills
Ability to understand new research results presented in original scientific papers.

## Learning goals: soft skills
Ability to present and to critically discuss these results in the framework of the corresponding area.

## Contents
Current conference and journal papers

## Prerequisites
none

## Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Seminar</td>
<td>10</td>
<td>2</td>
<td>30 T / 90 S</td>
<td>4</td>
</tr>
</tbody>
</table>

T = face-to-face teaching  
S = independent study

## Graded exams
Oral presentation, written report

## Ungraded coursework (required for admission to the exam)
MA-INF 3322  Applied Binary Exploitation

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>180 h</td>
<td>6 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

Module coordinator
Prof. Dr. Peter Martini

Lecturer(s)
Prof. Dr. Elmar Padilla

Programme
M. Sc. Computer Science
Mode
Optional
Semester
2. or 3.

Learning goals: technical skills
Static and dynamic program analysis, Exploitation (Stack-based Buffer Overflows, Format String Exploits, Heap Exploitation, Use-After-Free Exploits) and Countermeasures (Stack Cookies, NX, ASLR, RELRO)

Learning goals: soft skills
Frustration tolerance when working with binary representations and trying to apply taught techniques, focussed working on technically challenging problems, simultaneously applying knowledge from different areas of computer science

Contents
Our computers run a lot of closed source binary programs meaning that the source code of those programs is not available. Naturally, those programs contain bugs, mistakes that the programmer made during the development. Those bugs could (under certain circumstances) be exploited by attackers and thus may lead to arbitrary code execution. In this lecture we aim to teach you how to find well known exploitable bugs and how to exploit them.

After a brief recap of basic binary program analysis such as static and dynamic analysis, we will talk about vulnerability discovery in general, meaning that you will learn how to find exploitable bugs by yourself. Next we move on to basic stack-based buffer overflows and add mitigation techniques (stack cookies, NX, ASLR, RELRO, ...) as we progress and exploit them as well. After we finished the topic of stack-based buffer overflows we move on to more advanced topics such as heap exploitation, use-after-free exploits and others. The lecture ends with an introduction to fuzzing and an analysis of a sophisticated real-world exploit.

Prerequisites
Required:
none
Recommended:
• Binary Analysis skills (Lecture: “Applied Binary Analysis” BA-INF 155)
• Basic knowledge of the Linux operating system
• System Programming skills in C (Lecture: “Systemnahe Programmierung”)
• Basic Python programming skills

Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td>2</td>
<td>30 T / 45 S</td>
<td>2.5</td>
<td></td>
</tr>
<tr>
<td>Exercises</td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
<td></td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

Graded exams
Oral Examination

Ungraded coursework (required for admission to the exam)
The completion of regularly provided exercise sheets. The work can be done in groups of up to three students. A total of 50% of the points must be achieved. The exercises are divided into group tasks (four per exercise sheet) and tasks to be completed individually (one per exercise sheet) and the points to be achieved apply separately to both categories.

Literature
The relevant literature will be announced at the beginning of the lecture
MA-INF 3323  Lab Fuzzing Bootcamp

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

**Module coordinator**
Prof. Dr. Matthew Smith

**Lecturer(s)**
Dr. Christian Tiefenau

**Programme**
M. Sc. Computer Science

**Mode**
Optional

**Semester**
2. or 3.

**Learning goals: technical skills**
The students will carry out a practical task (project) in the context of fuzz testing, including test and documentation of the implemented software/system.

**Learning goals: soft skills**
Ability to properly present and defend design decisions, to prepare readable documentation of software; skills in constructively collaborating with others in small teams over a longer period of time; ability to classify one’s own results into the state-of-the-art of the resp. area.

**Contents**

**Prerequisites**
none

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lab</td>
<td>8</td>
<td>4</td>
<td>60 T / 210 S</td>
<td>9</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

**Graded exams**
Oral presentation, written report

**Ungraded coursework (required for admission to the exam)**
MA-INF 3324  Lab Design of Usable Security Mechanisms

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

Module coordinator
Prof. Dr. Matthew Smith

Lecturer(s)
Dr. Emmanuel von Zezschwitz

Programme
M. Sc. Computer Science

Mode
Optional

Semester
2. or 3.

Learning goals: technical skills
The students will carry out a practical task (project) in the context of usable security mechanisms, including test and documentation of the implemented software/system.

Learning goals: soft skills
Ability to properly present and defend design decisions, to prepare readable documentation of software; skills in constructively collaborating with others in small teams over a longer period of time; ability to classify ones own results into the state-of-the-art of the resp. area

Contents

Prerequisites
none

Teaching format

<table>
<thead>
<tr>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lab</td>
<td>8</td>
<td>4</td>
<td>9</td>
</tr>
</tbody>
</table>

Graded exams

Oral presentation, written report

Ungraded coursework (required for admission to the exam)
4 Intelligent Systems

MA-INF 4111 L2E2 6 CP Principles of Machine Learning ................................................... 105
MA-INF 4112 L2E2 6 CP Algorithms for Data Science ........................................................ 106
MA-INF 4113 L2E2 6 CP Cognitive Robotics .................................................................. 107
MA-INF 4114 L2E2 6 CP Robot Learning ................................................................. 108
MA-INF 4115 L3E1 6 CP Introduction to Natural Language Processing ......................... 109
MA-INF 4116 Sem2 4 CP AI Ethics Seminar ................................................................. 111
MA-INF 4201 L2E2 6 CP Artificial Life ........................................................................ 112
MA-INF 4203 L2E2 6 CP Autonomous Mobile Systems .................................................. 113
MA-INF 4204 L2E2 6 CP Technical Neural Nets ............................................................ 114
MA-INF 4208 Sem2 4 CP Seminar Vision Systems ....................................................... 115
MA-INF 4209 Sem2 4 CP Seminar Principles of Data Mining and Learning Algorithms ...... 116
MA-INF 4211 Sem2 4 CP Seminar Cognitive Robotics ................................................... 117
MA-INF 4213 Sem2 4 CP Seminar Humanoid Robots .................................................... 118
MA-INF 4214 Lab4 9 CP Lab Humanoid Robots ......................................................... 119
MA-INF 4215 L2E2 6 CP Humanoid Robotics ............................................................... 120
MA-INF 4216 L2E2 6 CP Biomedical Data Science & AI ............................................... 121
MA-INF 4217 Sem2 4 CP Seminar Machine Learning Methods in the Life Sciences ........ 122
MA-INF 4226 Lab4 9 CP Lab Parallel Computing for Mobile Robotics ............................ 123
MA-INF 4228 L4E2 9 CP Foundations of Data Science .................................................. 124
MA-INF 4230 L2E2 6 CP Advanced Methods of Information Retrieval ....................... 125
MA-INF 4231 Sem2 4 CP Seminar Advanced Topics in Information Retrieval .................. 126
MA-INF 4232 Lab4 9 CP Lab Information Retrieval in Practice ....................................... 127
MA-INF 4235 L2E2 6 CP Reinforcement Learning ....................................................... 128
MA-INF 4236 L2E2 4 CP Advanced Methods for Text Mining ....................................... 129
MA-INF 4237 Lab4 9 CP Natural Language Processing Lab .......................................... 131
MA-INF 4302 L2E2 6 CP Advanced Learning Systems .................................................. 132
MA-INF 4303 L2E2 6 CP Learning from Non-Standard Data ....................................... 133
MA-INF 4304 Lab4 9 CP Lab Cognitive Robotics .......................................................... 134
MA-INF 4306 Lab4 9 CP Lab Development and Application of Data Mining and Learning Systems 135
MA-INF 4308 Lab4 9 CP Lab Vision Systems ................................................................ 136
MA-INF 4309 Lab4 9 CP Lab Sensor Data Interpretation .............................................. 137
MA-INF 4310 Lab4 9 CP Lab Mobile Robots ............................................................... 138
MA-INF 4312 L2E2 6 CP Semantic Data Web Technologies .......................................... 139
MA-INF 4313 Sem2 4 CP Seminar Semantic Data Web Technologies ............................ 140
MA-INF 4314 Lab4 9 CP Lab Semantic Data Web Technologies .................................... 141
MA-INF 4316 L2E2 6 CP Graph Representation Learning ............................................... 142
MA-INF 4318 Sem2 4 CP Seminar Representation Learning for Big Data Analytics ........ 144
MA-INF 4322 L4E2 9 CP Lab Machine Learning on Encrypted Data .............................. 145
MA-INF 4324 Sem2 4 CP Seminar Advanced Topics in Data Science ............................ 146
MA-INF 4325 Lab4 9 CP Lab Data Science in Practice .................................................. 147
MA-INF 4326 L2E2 6 CP Explainable AI and Applications ............................................ 148
MA-INF 4327 Lab4 9 CP Lab Biomedical Data Science .................................................. 150
MA-INF 4328 L2E2 6 CP Spatio-Temporal Data Analytics ............................................. 151
MA-INF 4329 Sem2 4 CP Seminar Biological Intelligence ............................................... 152
MA-INF 4111 Principles of Machine Learning

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>180 h</td>
<td>6 CP</td>
<td>1 semester</td>
<td>every 2 years</td>
</tr>
</tbody>
</table>

Module coordinator: Prof. Dr.-Ing. Christian Bauckhage

Lecturer(s): Prof. Dr.-Ing. Christian Bauckhage

Programme: M. Sc. Computer Science

Mode: Optional


Learning goals: technical skills
Upon successful completion of this module, students should be able to describe fundamental methods, algorithms, and use cases of machine learning. Students acquire knowledge about supervised and unsupervised learning; based on the knowledge and skills acquired, students should be able to

- Implement, algorithms for optimization and parameter estimation in model training and machine learning tasks.
- Adopt the fundamental methods they learned about to a wide range of problems in automated intelligent data analysis.

Learning goals: soft skills
In the exercises, students can put their knowledge about theoretical concepts, mathematical methods, and algorithmic approaches into practice and realize small projects involving the implementation and evaluation of machine learning algorithms. This requires teamwork; upon successful completion of the module, students should be able to

- draft and implement basic machine learning algorithms for various practical problem settings
- prepare and give oral presentations about their work in front of an audience

Contents
Fundamental machine learning models for classification and clustering, model training via minimization of loss functions, fundamental optimization algorithms, model regularization, kernel methods for supervised and unsupervised learning, probabilistic modeling and inference, dimensionality reduction and latent factor models, the basic theory behind neural networks and neural network training; This course is intended to lay the foundation for more advanced courses on modern deep learning and reinforcement learning.

Prerequisites
Recommended:
Linear algebra, statistics, probability theory, calculus, python programming

Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td></td>
<td>2</td>
<td>30 T / 45 S</td>
<td>2.5</td>
</tr>
<tr>
<td>Exercises</td>
<td></td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

Graded exams
Written exam

Ungraded coursework (required for admission to the exam)
Successful exercise participation

Forms of media
- lecture slides / lecture notes are made available online
- notebooks with programming examples are made available online

Literature
• C.M. Bishop: Pattern Recognition and Machine Learning, Springer, 2006
• S. Haykin: Neural Networks and Learning Machines, Pearson, 2008
MA-INF 4112  Algorithms for Data Science

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>180 h</td>
<td>6 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

Module coordinator
Prof. Dr. Stefan Wrobel

Lecturer(s)
Dr. Tamas Horvath, Prof. Dr. Stefan Wrobel

Programme
M. Sc. Computer Science

Mode
Optional

Semester
1.

Learning goals: technical skills
In this module the students will learn algorithms for data science as well as implement and practice selected algorithms from this field. The module concentrates on basic algorithms in association rule mining, graph mining, and data streams. At the end of the module, students will be capable of analyzing formal properties of this kind of algorithms and choosing appropriate pattern discovery and data stream algorithms.

Learning goals: soft skills
Communicative skills (oral and written presentation of solutions, discussions in teams), self-competences (ability to accept and formulate criticism, ability to analyse, creativity in the context of an "open end" task), social skills (effective team work and project planning).

Contents
The module is offered every year, each time concentrating on one or more specific issues, such as frequent, closed and maximal frequent itemset mining, frequent subgraph mining algorithms for forests and for other graph classes beyond forests, frequent items and frequency moments in data streams, and graph stream algorithms.

Prerequisites
none

Course meetings

| Teaching format | Group size | h/week | Workload|h| CP |
|-----------------|------------|--------|----------|-----|
| Lecture         | 2          | 2      | 30 T / 45 S | 2.5 |
| Exercises       | 2          | 2      | 30 T / 75 S | 3.5 |

T = face-to-face teaching
S = independent study

Graded exams
Written exam

Ungraded coursework (required for admission to the exam)
Successful exercise participation

Forms of media
lectures, exercises

Literature
- Jiawei Han, Micheline Kamber, Jian Pei: Data Mining: Concepts and Techniques. Morgan Kaufmann Publishers, 2012.
MA-INF 4113  Cognitive Robotics

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>180 h</td>
<td>6 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

**Module coordinator**
Prof. Dr. Sven Behnke

**Lecturer(s)**
Prof. Dr. Sven Behnke

**Programme**
M. Sc. Computer Science

**Mode**
Optional

**Semester**
1. or 2.

**Learning goals: technical skills**
This lecture is one of two introductory lectures of the intelligent systems track. The lecture covers cognitive capabilities of robots, like self-localization, mapping, object perception, and action-planning in complex environments. This module complements MA-INF 4114 and can be taken before or after that module.

**Learning goals: soft skills**
Communicative skills (oral and written presentation of solutions, discussions in small teams), self competences (ability to accept and formulate criticism, ability to analyze problems)

**Contents**
Probabilistic approaches to state estimation (Bayes Filters, Kalman Filter, Particle Filter), motion models, sensor models, self-localization, mapping with known poses, simultaneous mapping and localization (SLAM), iterated closest-point matching, path planning, place- and person recognition, object recognition.

**Prerequisites**
Required:
MA-INF 4101 - Theory of Sensorimotor Systems has not been passed.

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td>2</td>
<td>2</td>
<td>30 T / 45 S</td>
<td>2.5</td>
</tr>
<tr>
<td>Exercises</td>
<td>2</td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

**Graded exams**
Written exam

**Ungraded coursework (required for admission to the exam)**
Successful exercise participation

**Literature**
MA-INF 4114  Robot Learning

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>180 h</td>
<td>6 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

Module coordinator
Prof. Dr. Sven Behnke

Lecturer(s)
Prof. Dr. Sven Behnke, Dr. Nils Goerke

Programme
M. Sc. Computer Science

Mode
Optional

Semester
1. or 2.

Learning goals: technical skills
This lecture is one of two introductory lectures of the intelligent systems track. Creating autonomous robots that can learn to assist humans in situations of daily life is a fascinating challenge for machine learning. The lecture covers key ingredients for a general robot learning approach to get closer towards human-like performance in robotics, such as reinforcement learning, learning models for control, learning motor primitives, learning from demonstrations and imitation learning, and interactive learning.

This module complements MA-INF 4113 and can be taken before or after that module.

Learning goals: soft skills
Communicative skills (oral and written presentation of solutions, discussions in small teams), self competences (ability to accept and formulate criticism, ability to analyze problems)

Contents
Reinforcement learning, Markov decision processes, dynamic programming, Monte Carlo methods, temporal-difference methods, function approximation, linear quadratic regulation, differential dynamic programming, partially observable MDPs, policy gradient methods, inverse reinforcement learning, imitation learning, learning kinematic models, perceiving and handling of objects.

Prerequisites
none

Course meetings
Teaching format
<table>
<thead>
<tr>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td>2</td>
<td>30 T / 45 S</td>
<td>2.5</td>
</tr>
<tr>
<td>Exercises</td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
</tr>
</tbody>
</table>

Graded exams
Written exam

Ungraded coursework (required for admission to the exam)
The completion of regularly provided exercise sheets. The work can be done in groups of up to two students. A total of 50% of the points must be achieved.

Literature
MA-INF 4115 Introduction to Natural Language Processing

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>180 h</td>
<td>6 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

**Module coordinator**
Prof. Dr. Lucie Flek

**Lecturer(s)**
Prof. Dr. Lucie Flek

**Programme**
M. Sc. Computer Science

**Mode**
Optional

**Semester**
1. or 2.

### Learning goals: technical skills
This class provides a technical perspective on NLP — methods for building computer software that understands and manipulates human language. Contemporary data-driven approaches are emphasized, focusing on machine learning techniques. The covered applications vary in complexity, including for example Entity Recognition, Argument Mining, or Emotion Analysis.

### Learning goals: soft skills
Group work during programming exercises will allow students to work on real-world NLP application projects. The final project offers you the chance to apply your newly acquired skills towards an in-depth application using different frameworks such as PyTorch and spaCy and present it in a poster session.

### Contents
Through lectures, exercises, and a final project, you will gain a thorough introduction to cutting-edge research in NLP, from the linguistic basis of computational language methods to recent advances in deep learning and large language models. This course provides:

- An overview of NLP goals, challenges, and applications
- Text representation (Words, sentences, paragraphs, documents), word embeddings, word2vec, BERT, word similarity
- Machine learning / deep learning algorithms for text classification, Transformers
- Basics of neural language modeling
- Basics of computational linguistics
- Transforming words to their base forms (tokenization, stemming, lemmatization)
- Syntactic analysis (part of speech tagging, chunking, and parsing)
- Techniques for extracting meaning from text (semantic analysis), use of lexical resources in NLP
- NLP applications and projects (e.g., Sentiment Analysis, Named Entity Recognition, Question Answering, Summarization, Fake news detection, Plagiarism detection, Abusive language detection, Opinion mining...)

### Prerequisites
**Recommended:**
- Basics of statistics recommended.
- Basic programming knowledge in Python is of advantage.
- Basics of machine learning are of advantage.

### Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td>3</td>
<td>45 T / 45 S</td>
<td>3</td>
<td></td>
</tr>
<tr>
<td>Exercises</td>
<td>1</td>
<td>15 T / 75 S</td>
<td>3</td>
<td></td>
</tr>
</tbody>
</table>

T = face-to-face teaching  
S = independent study

### Graded exams
Written exam (60 %); Project work (40 %)

### Ungraded coursework (required for admission to the exam)
Successful exercise participation
### Forms of media
- Lecture slides
- Exercise slides
- Notebooks with programming examples

### Literature
- J. Eisenstein: Introduction to Natural Language Processing
- Jurafsky, Daniel, and James H. Martin. "Speech and Language Processing: An Introduction to Natural Language Processing, Computational Linguistics, and Speech Recognition."
- S. Bird, E. Klein, E. Loper; Natural Language Processing with Python
MA-INF 4116  AI Ethics Seminar

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>120 h</td>
<td>4 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

Module coordinator
Prof. Dr. Lucie Flek

Lecturer(s)
Prof. Dr. Lucie Flek

Programme
M. Sc. Computer Science

Mode
Optional

Semester
1. or 2.

Learning goals: technical skills
The seminar aims to introduce students to the ethical dilemmas of artificial intelligence. Students will develop skills in assessing AI systems, identifying ethical dilemmas and social impacts, reasoning through ethical and social issues, and communicating their reasoning.

Learning goals: soft skills
Students will learn about the design of ethical and socially responsible systems. They will gain practice engaging with multidisciplinary perspectives from behavioral and social science. At the end of the course, students will write a final term essay on one of the course topics.

Contents
We study artificial intelligence and the ethical dilemmas associated with the research, design, deployment, and interaction with AI systems.

Six broad modules structure the seminar:
- Foundations of AI and AI ethics
- Bias & fairness
- Privacy & data privacy
- Social networks & civility of communication
- Politics & policy
- AI for “social good”

A typical lecture will consist of 2-3 student presentations that focus on a research article and the broad context of its topic.
Following each presentation, we discuss the work with a focus on assessing relevant ethical issues and potential approaches for ethical design and engineering.

Prerequisites
Required:
No previous knowledge is required.

Recommended:
Previously attended classes in machine learning, robotics, data mining, or related, can be useful for understanding the topics but are not a must.

Course meetings
Teaching format | Group size | h/week | Workload[h] | CP
Seminar | 10 | 2 | 30 T / 90 S | 4

T = face-to-face teaching
S = independent study

Graded exams
Written exam

Ungraded coursework (required for admission to the exam)
Successful exercise participation
MA-INF 4201 Artificial Life

Workload | Credit points | Duration | Frequency
--- | --- | --- | ---
180 h | 6 CP | 1 semester | every year

Module coordinator
Dr. Nils Goerke

Lecturer(s)
Dr. Nils Goerke

Programme
M. Sc. Computer Science

Mode
Optional

Semester
1-3.

Learning goals: technical skills
Detailed understanding of the most important approaches and principles of artificial life. Knowledge and understanding of the current state of research in the field of artificial life.

Learning goals: soft skills
Capability to identify the state of the art in artificial life, and to present and defend the found solutions within the exercises in front of a group of students. Critical discussion of the results of the homework.

Contents
Foundations of artificial life, cellular automata, Conway’s “Game of Life”; mechanisms for structural development; foundations of nonlinear dynamical systems, Lindenmeyer-systems, evolutionary methods and genetic algorithms, reinforcement learning, artificial immune systems, adaptive behaviour, self-organising criticality, multi-agent systems, and swarm intelligence, particle swarm optimization.

Prerequisites
none

Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td>2</td>
<td>30 T / 45 S</td>
<td>2.5</td>
<td></td>
</tr>
<tr>
<td>Exercises</td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
<td></td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

Graded exams
Written exam

Ungraded coursework (required for admission to the exam)
The completion of regularly provided exercise sheets. The work can be done in groups of up to two students. A total of 50% of the points must be achieved. Each student must present a solution to an exercise in the exercise sessions twice.

Forms of media
Pencil and paper work, explain solutions in front of the exercise group, implementation of small programs, use of simple simulation tools.

Literature
- Christoph Adami: Introduction to Artificial Life, The Electronic Library of Science, TELOS, Springer-Verlag
MA-INF 4203  Autonomous Mobile Systems

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>180 h</td>
<td>6 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

**Module coordinator**
Prof. Dr. Sven Behnke

**Lecturer(s)**
Dr. Dirk Schulz, Prof. Dr. Sven Behnke

**Programme**
M. Sc. Computer Science

**Mode**
Optional

**Semester**
2.

**Learning goals: technical skills**
Profound knowledge of development and test regarding structure and function of learning, autonomous, mobile systems; Knowledge of the computational, mathematical, and technical requirements for the design of autonomous systems for specific applications and for specific functional environments.

**Learning goals: soft skills**
The students will be capable to assess applications for autonomous mobile systems. They will be capable to identify what part of the applications might be improved by using state of the art developments. The student will learn how to plan and implement a software project in small working groups.

**Contents**
Requirements for the implementation of autonomous mobile systems, e.g. for: map making, dead reckoning, localisation, SLAM-methods, various principles of robot path planning; methods for action planning. Comparison of different learning paradigms for specific applications.

**Prerequisites**
Recommended:
any of the following:
MA-INF 4101 – Theory of Sensorimotor Systems
MA-INF 4113 – Cognitive Robotics

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td>2</td>
<td>30 T / 45 S</td>
<td>2.5</td>
<td></td>
</tr>
<tr>
<td>Exercises</td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
<td></td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

**Graded exams**
Oral exam

**Ungraded coursework (required for admission to the exam)**
Successful exercise participation

**Literature**
- J. Buchli: Mobile Robots: Moving Intelligence, Published by Advanced Robotic Systems and Pro Literatur Verlag
- Howie Choset et al.: Principles of Robot Motion, MIT-Press, 2005
MA-INF 4204  Technical Neural Nets

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>180 h</td>
<td>6 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

Module coordinator: Dr. Nils Goerke
Lecturer(s): Dr. Nils Goerke

Programme: M. Sc. Computer Science
Mode: Optional

Learning goals: technical skills
Detailed knowledge of the most important neural network approaches and learning algorithms and its fields of application. Knowledge and understanding of technical neural networks as Non-Von Neumann computer architectures similar to concepts of brain functions at different stages of development.

Learning goals: soft skills
The students will be capable to propose several paradigms from neural networks that are capable to solve a given task. They can discuss the pro and cons with respect to efficiency and risk. They will be capable to plan and implement a small project with state of the art neural network solutions.

Contents
Multi-layer perceptron, radial-basis function nets, Hopfield nets, self organizing maps (Kohonen), adaptive resonance theory, learning vector quantization, recurrent networks, back-propagation of error, reinforcement learning, Q-learning, support vector machines, pulse processing neural networks. Exemplary applications of neural nets: function approximation, prediction, quality control, image processing, speech processing, action planning, control of technical processes and robots. Implementation of neural networks in hardware and software: tools, simulators, analog and digital neural hardware.

Prerequisites
none

Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td>2</td>
<td></td>
<td>30 T / 45 S</td>
<td>2.5</td>
</tr>
<tr>
<td>Exercises</td>
<td>2</td>
<td></td>
<td>30 T / 75 S</td>
<td>3.5</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

Graded exams
Written exam

Ungraded coursework (required for admission to the exam)
Successful exercise participation

Forms of media
Pencil and paper work, explaining solutions in front of the exercise group, implementation of small programs, use of simple simulation tools

Literature
MA-INF 4208  Seminar Vision Systems

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>120 h</td>
<td>4 CP</td>
<td>1 semester</td>
<td>every semester</td>
</tr>
</tbody>
</table>

Module coordinator
Prof. Dr. Sven Behnke

Lecturer(s)
Prof. Dr. Sven Behnke, Prof. Dr. Joachim K. Anlauf, Dr. Nils Goerke

Programme
M. Sc. Computer Science

Mode
Optional

Semester
2. or 3.

Learning goals: technical skills
- Knowledge in advanced topics in the area of technical vision systems, such as image segmentation, feature extraction, and object recognition.
- Ability to understand new research results presented in original scientific papers and to present them in a research talk as well as in a seminar report.

Learning goals: soft skills
Self-competences (time management, literature search, self-study), communication skills (preparation and clear didactic presentation of research talk, scientific discussion, structured writing of seminar report), social skills (ability to formulate and accept criticism, critical examination of research results).

Contents
Current research papers from conferences and journals in the field of vision systems covering fundamental techniques and applications.

Prerequisites
Recommended:
At least 1 of the following:
MA-INF 4111 – Intelligent Learning and Analysis Systems: Machine Learning
MA-INF 4204 – Technical Neural Nets

Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Seminar</td>
<td>10</td>
<td>2</td>
<td>30 T / 90 S</td>
<td>4</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

Graded exams
Oral presentation, written report

Ungraded coursework (required for admission to the exam)

Literature
MA-INF 4209 Seminar Principles of Data Mining and Learning Algorithms

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>120 h</td>
<td>4 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

Module coordinator
Prof. Dr. Stefan Wrobel

Lecturer(s)
Prof. Dr. Stefan Wrobel

Programme
M. Sc. Computer Science

Mode
Optional

Semester
2. or 3.

Learning goals: technical skills
Enhanced and in-depth knowledge in specialized topics in the area of machine learning and data mining, acquiring the competence to independently study scientific literature, present it to others and discuss it with a knowledgeable scientific auditorium. Learn how to scientifically present prior work by others, in writing and in presentations.

Learning goals: soft skills
Communicative skills (preparing and presenting talks, written presentation of contents in a longer document), self competences (time management with long-ranging deadlines, ability to accept and formulate criticism, ability to analyse, creativity).

Contents
Theoretical, statistical and algorithmical principles of data mining and learning algorithms. Search and optimization algorithms. Specialized learning algorithms from the frontier of research. Fundamental results from neighbouring areas.

Prerequisites
Recommended:
At least 1 of the following:
MA-INF 4111 – Intelligent Learning and Analysis Systems: Machine Learning
MA-INF 4112 – Intelligent Learning and Analysis Systems: Data Mining and Knowledge Discovery

Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Seminar</td>
<td>10</td>
<td>2</td>
<td>30 T / 90 S</td>
<td>4</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

Graded exams
Oral presentation, written report

Ungraded coursework (required for admission to the exam)

Forms of media
Scientific papers and websites, interactive presentations.

Literature
The relevant literature will be announced towards the end of the previous semester.
MA-INF 4211  Seminar Cognitive Robotics

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>120 h</td>
<td>4 CP</td>
<td>1 semester</td>
<td>every semester</td>
</tr>
</tbody>
</table>

**Module coordinator**

Prof. Dr. Sven Behnke

**Lecturer(s)**

Prof. Dr. Sven Behnke, Dr. Nils Goerke

**Programme**

M. Sc. Computer Science

**Mode**

Optional

**Semester**

2. or 3.

**Learning goals: technical skills**

Knowledge in advanced topics in the area of cognitive robotics, such as robot perception, action planning, and robot learning.

Ability to understand new research results presented in original scientific papers and to present them in a research talk as well as in a seminar report.

**Learning goals: soft skills**

Self-competences (time management, literature search, self-study), communication skills (preparation and clear didactic presentation of research talk, scientific discussion, structured writing of seminar report), social skills (ability to formulate and accept criticism, critical examination of research results).

**Contents**

Current research papers from conferences and journals in the field of cognitive robotics covering fundamental techniques and applications.

**Prerequisites**

**Recommended:**

At least 1 of the following:

MA-INF 4113 – Cognitive Robotics

MA-INF 4114 – Robot Learning

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Seminar</td>
<td>10</td>
<td>2</td>
<td>30 T / 90 S</td>
<td>4</td>
</tr>
</tbody>
</table>

T = face-to-face teaching

S = independent study

**Graded exams**

Oral presentation, written report

**Ungraded coursework (required for admission to the exam)**

**Literature**

- Selected papers.
**MA-INF 4213  Seminar Humanoid Robots**

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>120 h</td>
<td>4 CP</td>
<td>1 semester</td>
<td>every semester</td>
</tr>
</tbody>
</table>

**Module coordinator**
Prof. Dr. Maren Bennewitz

**Lecturer(s)**
Prof. Dr. Maren Bennewitz

**Programme**
M. Sc. Computer Science

**Mode**
Optional

**Semester**
2.

**Learning goals: technical skills**
Knowledge in advanced topics in the area of humanoid robotics, such as environment perception, state estimation, navigation, or motion planning. Ability to understand new research results of scientific papers and to present them in a talk as well as in a self-written summary.

**Learning goals: soft skills**
Self-competences (time management, literature search, self-study), communication skills (preparation of the talk, clear didactic presentation of techniques and experimental results, scientific discussion, structured writing of summary), social skills (ability to formulate and accept criticism, critical examination of algorithms and experimental results).

**Contents**
Current research papers from conferences and journals in the field of humanoid robotics covering fundamental techniques and applications.

**Prerequisites**
**Recommended:**
At least 1 of the following:
MA-INF 4215 – Humanoid Robotics
MA-INF 4113 – Cognitive Robotics

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Seminar</td>
<td>10</td>
<td>2</td>
<td>30 T / 90 S</td>
<td>4</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

**Graded exams**
Oral presentation, written report

**Ungraded coursework (required for admission to the exam)**

**Literature**
- B. Siciliano, O. Khatib (Eds.): Springer Handbook of Robotics
- K. Harada, E. Yoshida, K. Yokoi (Eds.), Motion Planning for Humanoid Robots, Springer
- Selected papers.
MA-INF 4214  Lab Humanoid Robots

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>every semester</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Module coordinator</th>
<th>Lecturer(s)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Prof. Dr. Maren Bennewitz</td>
<td>Prof. Dr. Maren Bennewitz</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Programme</th>
<th>Mode</th>
<th>Semester</th>
</tr>
</thead>
<tbody>
<tr>
<td>M. Sc. Computer Science</td>
<td>Optional</td>
<td>2.</td>
</tr>
</tbody>
</table>

**Learning goals: technical skills**
Practical experience and in-depth knowledge in the design and implementation of perception, state estimation, environment representation, navigation, and motion planning techniques for humanoid robots. In small groups, the participants analyze a problem, realize a solution, and perform an experimental evaluation.

**Learning goals: soft skills**
Self-competences (time management, goal-oriented work, ability to analyze problems theoretically and to find practical solutions), communication skills (collaboration in small teams, oral and written presentation of solutions, critical examination of implementations).

**Contents**
Robot middleware, perception, state estimation, environment representations, navigation, and motion planning for humanoid robots.

**Prerequisites**
**Recommended:**
At least 1 of the following:
MA-INF 4215 – Humanoid Robotics
MA-INF 4113 – Cognitive Robotics

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lab</td>
<td>8</td>
<td>4</td>
<td>60 T / 210 S</td>
<td>9</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

**Graded exams**
Oral presentation, written report

**Ungraded coursework (required for admission to the exam)**

**Literature**
- B. Siciliano, O. Khatib (Eds.): Springer Handbook of Robotics
- K. Harada, E. Yoshida, K. Yokoi (Eds.), Motion Planning for Humanoid Robots, Springer
- Selected papers.
MA-INF 4215  Humanoid Robotics

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>180 h</td>
<td>6 CP</td>
<td>1 semester</td>
<td>at least every 2 years</td>
</tr>
</tbody>
</table>

**Module coordinator**
Prof. Dr. Maren Bennewitz

**Lecturer(s)**
Prof. Dr. Maren Bennewitz

**Programme**
M. Sc. Computer Science

**Mode**
Optional

**Semester**
2-3.

**Learning goals: technical skills**
This lecture covers techniques for humanoid robots such as perception, navigation, and motion planning.

**Learning goals: soft skills**
Communicative skills (oral and written presentation of solutions, discussions in small teams), ability to analyze problems.

**Contents**
Self-calibration with least squares, 3D environment representations, self-localization with particle filters, footstep planning, inverse kinematics, whole-body motion planning with rapidly exploring random trees, statistical testing.

**Prerequisites**

**Recommended:**
MA-INF 4113 – Cognitive Robotics

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td>2</td>
<td>2</td>
<td>30 T / 45 S</td>
<td>2.5</td>
</tr>
<tr>
<td>Exercises</td>
<td>2</td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

**Graded exams**

Oral exam

**Ungraded coursework (required for admission to the exam)**
Successful exercise participation

**Literature**
- B. Siciliano, O. Khatib (Eds.): Springer Handbook of Robotics
- K. Harada, E. Yoshida, K. Yokoi (Eds.), Motion Planning for Humanoid Robots, Springer
- Selected research papers.
MA-INF 4216  Biomedical Data Science & AI

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>180 h</td>
<td>6 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

Module coordinator
Dr. Holger Fröhlich

Lecturer(s)
Dr. Holger Fröhlich

Programme
M. Sc. Computer Science

Mode
Optional

Semester
3.

Learning goals: technical skills
- understanding and knowledge of fundamental data mining and machine learning methods
- understanding of their application in bioinformatics

Learning goals: soft skills
- communication: oral and written presentation of solutions to exercises
- self-competences: ability to analyze application problems and to formulate possible solutions
- practical skills: ability to practically implement solutions
- social skills: working in a small team with other students

Contents
This lecture gives a broad overview about frequently used statistical techniques as well as data mining and machine learning algorithms. The use of the respective methods to solve problems in bioinformatics is explained. The goal is to understand the explained methods, being able to apply them correctly and partially implement them. More detailed, the following topics are covered in the context of their application in bioinformatics:
- Short introduction to Bioinformatics and Biomedicine
- Statistical Basics: Probability distributions and Bayesian inference, statistical hypothesis testing, linear models, logistic regression, Principal Component Analysis
- Clustering
- Hidden Markov Models
- Principles of Supervised Machine Learning
- Elastic Net
- Basics of deep learning

Prerequisites
none

Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload/h</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td>2</td>
<td>30 T / 45 S</td>
<td>2.5</td>
<td></td>
</tr>
<tr>
<td>Exercises</td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
<td></td>
</tr>
</tbody>
</table>

Graded exams

Written exam

Ungraded coursework (required for admission to the exam)
(i) The completion of regularly provided exercise sheets. The work can be done in groups of up to three students. A total of 50% of the points must be achieved. Each student must present a solution to an exercise in the exercise sessions once. (ii) Participation in an achievement test. On the test, at least 50% of the points much be achieved.

Literature
S. Boslaugh, P. Watters, Statistics in a Nutshell, O’Reilly, 2008
N. Jones, P. Pevzner, An Introduction to Bioinformatics Algorithms, MIT Press, 2004
MA-INF 4217  Seminar Machine Learning Methods in the Life Sciences

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>120 h</td>
<td>4 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

**Module coordinator**
Dr. Holger Fröhlich

**Lecturer(s)**
Dr. Holger Fröhlich

**Programme**
M. Sc. Computer Science

**Mode**
Optional

**Semester**
4.

**Learning goals: technical skills**
- understanding and knowledge of machine learning methods and their application in modern life sciences, e.g. biomedicine

**Learning goals: soft skills**
- communication: oral scientific presentation of a defined topic
- self-competences: ability to identify relevant literature for a given topic; ability to read, understand and analyze scientific publications
- social skills: ability to discuss a scientific topic with other students and the staff

**Contents**
Machine learning techniques play a crucial role in modern life sciences, including biomedicine. The goal of this seminar is to discuss a variety of machine learning techniques in the context of their application to solve real-world problems in biomedicine.

Topics will be selected from the following areas:
- Ensemble learning
- Survival and disease progression models
- Bayesian Networks
- Stochastic processes, e.g. Gaussian Processes, Dirichlet Process Mixture Models
- MCMC methods
- Deep learning methods, e.g. DNNs, CNNs, Deep Belief Networks
- feature selection and non-linear embedding methods
- multi-modal data fusion techniques

Attendees will be asked to perform research about their topic in a self-responsible manner.

**Prerequisites**
**Recommended:**
MA-INF 4216 – Data Mining and Machine Learning Methods in Bioinformatics

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
<th>T = face-to-face teaching</th>
<th>S = independent study</th>
</tr>
</thead>
<tbody>
<tr>
<td>Seminar</td>
<td>10</td>
<td>2</td>
<td>30 T / 90 S</td>
<td>4</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

**Graded exams**
Oral presentation, written report

**Ungraded coursework (required for admission to the exam)**

**Forms of media**
powerpoint

**Literature**
selected journal and conference papers
**MA-INF 4226  Lab Parallel Computing for Mobile Robotics**

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>at least every 2 years</td>
</tr>
</tbody>
</table>

**Module coordinator**
Prof. Dr. Maren Bennewitz

**Lecturer(s)**
Prof. Dr. Maren Bennewitz

**Programme**
M. Sc. Computer Science

**Mode**
Optional

**Semester**
2.

**Learning goals: technical skills**
Students will make practical experience with the design and implementation of parallelized algorithms in the context of motion planning and navigation.

**Learning goals: soft skills**
Ability to properly present and defend design decisions, to prepare readable documentation of software; skills in constructively collaborating with others in small teams over a longer period of time; ability to classify ones own results into the state-of-the-art of the resp. area

**Contents**
Parallel programming on the GPU, CUDA, shortest path planning, collision checking, visibility graph, A* algorithm

**Prerequisites**
Recommended:
C++, Linux.

Since the exercises revolve around path planning, one of those courses might be helpful:
MA-INF 4203: Autonomous Mobile Systems
MA-INF 4113: Cognitive Robotics
MA-INF 4310: Lab Mobile Robots

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lab</td>
<td>8</td>
<td>4</td>
<td>60 T / 210 S</td>
<td>9</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

**Graded exams**
Oral presentation, written report

**Ungraded coursework (required for admission to the exam)**
MA-INF 4228  Foundations of Data Science

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Module coordinator</th>
<th>Lecturer(s)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Dr. Michael Nüsken</td>
<td>Dr. Michael Nüsken</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Programme</th>
<th>Mode</th>
<th>Semester</th>
</tr>
</thead>
<tbody>
<tr>
<td>M. Sc. Computer Science</td>
<td>Optional</td>
<td>2. or 3.</td>
</tr>
</tbody>
</table>

**Learning goals: technical skills**


Skills: Understanding of mathematical tools.

**Learning goals: soft skills**

Competences: Application to data science problems and ability to assess similar methods.

**Contents**

Data science aims at making sense of big data. To that end, various tools have to be understood for helping in analyzing the arising structures.

Often data comes as a collection of vectors with a large number of components. To understand their common structure is the first main objective of understanding the data. The geometry and the linear algebra behind them becomes relevant and enlightening. Yet, the intuition from low-dimensional space turns out to be often misleading. We need to be aware of the particular properties of high-dimensional spaces when working with such data. Fruitful methods for the analysis include singular vector decomposition from linear algebra and supervised and unsupervised machine learning. If time permits, we also consider random graphs, which are the second most used model for real world phenomena.

**Prerequisites**

none

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td>4</td>
<td>60 T / 105 S</td>
<td>5.5</td>
<td></td>
</tr>
<tr>
<td>Exercises</td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
<td></td>
</tr>
</tbody>
</table>

**Graded exams**

Written exam

**Ungraded coursework (required for admission to the exam)**

The completion of regularly provided exercise sheets. A total of 50% of the points must be achieved. Each student must present a solution to an exercise in the exercise sessions twice.

**Literature**

MA-INF 4230  Advanced Methods of Information Retrieval

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>180 h</td>
<td>6 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

Module coordinator

Prof. Dr. Elena Demidova

Lecturer(s)

Prof. Dr. Elena Demidova

Programme

M. Sc. Computer Science

Mode

Optional

Semester

2. or 3.

Learning goals: technical skills

This module introduces the students to the advanced methods, data structures, and algorithms of information retrieval for structured and semi-structured data (including, for example, knowledge graphs, relational data, and tabular data).

At the end of the module, the students will be capable of choosing appropriate data structures and retrieval algorithms for specific applications and correctly apply relevant statistical and machine learning-based information retrieval procedures.

Learning goals: soft skills

Communication skills: oral and written presentation and discussion of solutions.

Self-competences: ability to analyse and solve problems.

Contents

The module topics include data structures, ranking methods, and efficient algorithms that enable end-users to effectively obtain the most relevant search results from structured, heterogeneous, and distributed data sources. Furthermore, we will study the corresponding evaluation techniques as well as novel applications.

Prerequisites

none

Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td>2</td>
<td>30 T / 45 S</td>
<td>2.5</td>
<td></td>
</tr>
<tr>
<td>Exercises</td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
<td></td>
</tr>
</tbody>
</table>

T = face-to-face teaching

S = independent study

Graded exams

Written exam

Ungraded coursework (required for admission to the exam)

The completion of regularly provided exercise sheets. The work can be done in groups of up to three, four or five students, depending on the total number of students taking the course. A total of 50% of the points must be achieved. For 80% of the exercise sheets, 40% of the points must be achieved for each sheet. Each student must present a solution to an exercise in the exercise sessions once.

Literature

Selected chapters from:


Further references to relevant material will be provided during the lecture.
MA-INF 4231  Seminar Advanced Topics in Information Retrieval

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>120 h</td>
<td>4 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

Module coordinator
Prof. Dr. Elena Demidova

Lecturer(s)
Prof. Dr. Elena Demidova

Programme
M. Sc. Computer Science

Mode
Optional

Semester
2. or 3.

Learning goals: technical skills
This module concentrates on specialized topics in information retrieval. The students obtain skills in the independent, in-depth study of state-of-the-art scientific literature on specific topics, discussion with their peers and presentation to the scientific audience.

Learning goals: soft skills
Communication skills: oral and written presentation of scientific content. Self-competences: the ability to analyze problems, time management, creativity.

Contents
Statistical and machine learning-based information retrieval methods, including typical steps of the information retrieval process: data collection, feature extraction, indexing, retrieval, ranking, and evaluation. Specialized data representation and retrieval methods for selected data types and applications in specific domains.

Prerequisites
Recommended: MA-INF 4230 - Advanced Methods of Information Retrieval

Course meetings
<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Seminar</td>
<td>10</td>
<td>2</td>
<td>30 T / 90 S</td>
<td>4</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

Graded exams
Oral presentation, written report

Ungraded coursework (required for admission to the exam)
None

Literature
Selected chapters from:

Further relevant literature will be announced at the beginning of the seminar.
MA-INF 4232  Lab Information Retrieval in Practice

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

**Module coordinator**
Prof. Dr. Elena Demidova
Lecturer(s)
Prof. Dr. Elena Demidova

**Programme**
M. Sc. Computer Science

**Mode**
Optional

**Semester**
2. or 3.

**Learning goals: technical skills**
This module concentrates on practical experience in information retrieval. Participants acquire basic knowledge and practical experience in designing and implementing information retrieval systems for specific data types and applications.

**Learning goals: soft skills**
Communication skills: the ability to work in teams.
Self-competences: the ability to analyse problems and find practical solutions. Time management, creativity, presentation of results.

**Contents**
Practical application of information retrieval methods to solve retrieval problems on real-world data and evaluate proposed solutions.

**Prerequisites**
Recommended:
MA-INF 4230 - Advanced Methods of Information Retrieval
MA-INF 4231 - Seminar Advanced Topics in Information Retrieval

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lab</td>
<td>8</td>
<td>4</td>
<td>60 T / 210 S</td>
<td>9</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

**Graded exams**
Oral presentation, written report

**Ungraded coursework (required for admission to the exam)**
None

**Literature**
Selected chapters from:


Further references to relevant material will be provided during the lab.
MA-INF 4235  Reinforcement Learning

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>180 h</td>
<td>6 CP</td>
<td>1 semester</td>
<td>every 2 years</td>
</tr>
</tbody>
</table>

**Module coordinator**

Prof. Dr.-Ing. Christian Bauckhage

**Lecturer(s)**

Prof. Dr.-Ing. Christian Bauckhage

**Programme**

M. Sc. Computer Science

**Mode**

Optional

**Semester**

2-3.

**Learning goals: technical skills**

Upon successful completion of this module, students should be able to describe fundamental methods, algorithms, and use cases of reinforcement learning. Students acquire knowledge about underlying mathematical models and corresponding algorithms; based on the knowledge and skills acquired, students should be able to:

- implement algorithms for reinforcement learning problems;
- adopt the fundamental methods they learned about to a wide range of problems in policy optimization.

**Learning goals: soft skills**

In the exercises, students can put their knowledge about theoretical concepts, mathematical methods, and algorithmic approaches into practice and realize small projects involving the implementation and evaluation of search- and policy learning algorithms. This requires teamwork; upon successful completion of the module, students should be able to:

- draft and implement basic reward functions and policy learning algorithms for various practical problem settings;
- prepare and give oral presentations about their work in front of an audience.

**Contents**

State space models, tree search algorithms, Monte Carlo tree search, Markov chain models, Markov decision processes, value functions, reward functions, Bellman equations, policy learning, TD learning Q learning, deep Q learning

**Prerequisites**

Required:

- Linear algebra, statistics, probability theory, python programming

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td></td>
<td>2</td>
<td>30 T / 45 S</td>
<td>2.5</td>
</tr>
<tr>
<td>Exercises</td>
<td></td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
</tr>
</tbody>
</table>

T = face-to-face teaching

S = independent study

**Graded exams**

Written exam

**Ungraded coursework (required for admission to the exam)**

The completion of regularly provided exercise sheets. The work can be done in groups of up to four students. A total of 50% of the points must be achieved. Each student must present a solution to an exercise in the exercise sessions once.

**Forms of media**

- lecture slides / lecture notes are made available online
- notebooks with programming examples are made available online

**Literature**

R.S. Sutton and A.G. Barto: Reinforcement Learning, 2nd ed., MIT Press, 2018
MA-INF 4236  Advanced Methods for Text Mining

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>120 h</td>
<td>4 CP</td>
<td>1 semester</td>
<td>at least every 2 years</td>
</tr>
</tbody>
</table>

Module coordinator
Prof. Dr. Rafet Sifa

Programme
M. Sc. Computer Science

Mode
Optional

Semester
2-4.

Learning goals: technical skills
Knowledge: Students will learn about the basic as well as the advanced methods for processing textual data, including necessary preprocessing steps such as stemming and lemmatization. They will also learn about representation learning methods, such as TF-IDF, Latent Semantic Indexing, Global Vectors, Recurrent Neural Networks, Transformer Networks, as well as the variants of the last such as Generative Pre-trained Transformers and Bidirectional Encoder Representations from Transformers, to extract meaningful embeddings for downstream tasks. The students will gain knowledge on how to build predictive and prescriptive methods for a variety of objectives, including text classification, outlier detection, and recommender systems. Additionally, they will learn how to categorize these methods based on their complexities and their applicability to different text mining problems, such as sentiment analysis, natural language inference, computational argumentation, information extraction, named entity recognition, text summarization, opinion mining, text segmentation, event detection, and more.

Skill: Students should be able to analyze, design as well as reason about existing and new data mining algorithms, theoretically compare algorithms, strengthen their analytical thinking to solve difficult modelling problems, have acquired the necessary mathematical as well as programming/IT skills to systematically plan, design and implement text and data mining projects.

Competences: Based on the knowledge and skills acquired in this module, the students will be able to assess certain characteristics of the already existing text mining methods as well as build new solutions to emerging problems. Additionally, the students will be able to transfer their knowledge to other data science areas involving modelling data with sequential dependencies.

Learning goals: soft skills

Contents
Neural Networks, Text Mining Pipelines, Stemming, Lemmatization, TF-IDF, Latent Semantic Indexing, Global Vectors, Recurrent Neural Networks, Transformer Networks, Generative Pre-trained Transformers, Bidirectional Encoder Representations, Prompt Analysis, Sentiment Analysis, Natural Language Inference, Computational Argumentation, Information Extraction, Named Entity Recognition, Text Summarization, Opinion Mining, Text Segmentation, Event Detection, Representation Learning and Applications

Prerequisites
Recommended:
Basic knowledge of AI, data science, machine learning, and pattern recognition; programming skills; good working knowledge in statistics, linear algebra, and optimization.

Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td></td>
<td>2</td>
<td>30 T / 45 S</td>
<td>2.5</td>
</tr>
<tr>
<td>Exercises</td>
<td></td>
<td>1</td>
<td>15 T / 30 S</td>
<td>1.5</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

Graded exams
Written exam

Ungraded coursework (required for admission to the exam)
Successful exercise participation (written homework as well as the given programming assignments)
Literature

- Introduction to Information Retrieval, Christopher D. Manning, Prabhakar Raghavan and Heinrich Schütze
- Lecture notes of the instructors
MA-INF 4237  Natural Language Processing Lab

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

**Module coordinator**
Prof. Dr. Lucie Flek

**Lecturer(s)**
Prof. Dr. Lucie Flek

**Programme**
M. Sc. Computer Science

**Mode**
Optional

**Semester**
2-3.

Learning goals: technical skills
The Natural Language Processing (NLP) Lab course provides students with a detailed look at the recent advancements in NLP, covering various aspects such as large language models (LLMs), conversational systems, and computational social science. The course emphasizes a practical approach and offers you the opportunity to gain hands-on experience in developing NLP-based systems, allowing you to deepen your understanding of NLP technologies and apply theoretical knowledge to real-world scenarios.

Learning goals: soft skills
Through tutorials and a final project, you will gain practical skills in NLP techniques and have this chance to apply this knowledge to a various interesting project. Students will collaborate in small teams (a group of two students) and implement NLP applications over the course of the term. Each team is advised by one researcher of the CAISA Lab.

Contents
The course emphasizes a practical approach and offers you the opportunity to gain hands-on experience in developing NLP-based systems, allowing you to deepen your understanding of NLP technologies and apply theoretical knowledge to real-world scenarios.

Prerequisites
**Required:**
MA-INF 4115: Introduction to Natural Language Processing

**Recommended:**
- Basic programming knowledge in Python and Machine Learning
- Basics of Machine Learning
- Basic knowledge of Python Libraries for ML (NumPy, Scikit-Learn, Pandas)
- Basics of Probability, Linear Algebra and Statistics

Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lab</td>
<td>8</td>
<td>4</td>
<td>60 T / 210 S</td>
<td>9</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

Graded exams
Oral presentation, written report

Ungraded coursework (required for admission to the exam)
Successful exercise participation
MA-INF 4302  Advanced Learning Systems

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>180 h</td>
<td>6 CP</td>
<td>1 semester</td>
<td>every 2 years</td>
</tr>
</tbody>
</table>

Module coordinator
Prof. Dr. Stefan Wrobel

Lecturer(s)
Prof. Dr. Stefan Wrobel

Programme
M. Sc. Computer Science

Mode
Optional

Semester
2. or 3.

Learning goals: technical skills
Participants specialize and require in-depth knowledge of one particular class of learning algorithms, they acquire the necessary knowledge to improve existing algorithms and construct their own within the given class, all the way up to the research frontier on the topic.

Learning goals: soft skills
In group work, students acquire the necessary social and communication skills for effective team work and project planning, and learn how to present software projects to others.

Contents
The module each time concentrates on one or more specific algorithm classes, e.g.

- kernel machines
- neural networks
- probabilistic and statistical learning approaches
- logic-based learning approaches
- reinforcement learning

Prerequisites
Recommended:
all of the following:
MA-INF 4111 – Intelligent Learning and Analysis Systems: Machine Learning
MA-INF 4112 – Intelligent Learning and Analysis Systems: Data Mining and Knowledge Discovery

Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td>2</td>
<td>30 T / 45 S</td>
<td>2.5</td>
<td></td>
</tr>
<tr>
<td>Exercises</td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
<td></td>
</tr>
</tbody>
</table>

Graded exams
Written exam

Ungraded coursework (required for admission to the exam)
Successful exercise participation

Forms of media
lectures, exercises, software systems

Literature
- John Shawe-Taylor, Nello Christianini, Kernel Methods for Pattern Analysis, CUP, 2004
- Christopher Bishop, Pattern Recognition and Machine Learning, The University of Edinburgh, 2006
- Richard Duda, Peter Hart, David Stork, Pattern Classification, John Wiley and Sons, 2001
MA-INF 4303  Learning from Non-Standard Data

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>180 h</td>
<td>6 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

**Module coordinator**
Prof. Dr. Stefan Wrobel

**Lecturer(s)**
Prof. Dr. Stefan Wrobel, Dr. Tamas Horvath

**Programme**
M. Sc. Computer Science

**Mode**
Optional

**Semester**
2. or 3.

---

**Learning goals: technical skills**

Participants deepen their knowledge of learning systems with respect to one particular non-standard data type, i.e., non-tabular data, as they are becoming increasingly important in many applications. Each type of data not only requires specialized algorithms but also knowledge of the surrounding pre- and postprocessing operations which is acquired by the participants in the module. In group work, students acquire the necessary social and communication skills for effective team work and project planning, and learn how to present software projects to others.

**Learning goals: soft skills**

Communicative skills (oral and written presentation of solutions, discussions in teams), self-competences (ability to accept and formulate criticism, ability to analyse, creativity in the context of an 'open end' task)

---

**Contents**

The module will offered every year, concentrating on one particular non-standard data type each time, including: Text Mining, Multimedia Mining, Graph Mining. Learning from structured data, Spatial Data Mining

---

**Prerequisites**

Recommended:
all of the following:

- MA-INF 4111 – Intelligent Learning and Analysis Systems: Machine Learning
- MA-INF 4112 – Intelligent Learning and Analysis Systems: Data Mining and Knowledge Discovery

---

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload/h</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td></td>
<td></td>
<td>30 T / 45 S</td>
<td>2.5</td>
</tr>
<tr>
<td>Exercises</td>
<td></td>
<td></td>
<td>30 T / 75 S</td>
<td>3.5</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

---

**Graded exams**

Written exam

---

**Ungraded coursework (required for admission to the exam)**

Successful exercise participation

---

**Forms of media**

lectures, exercises, software systems.

---

**Literature**

- Diane J. Cook, Lawrence B. Holder, Mining Graph Data, Wiley & Sons, 2006
- Saso Dzeroski, Nada Lavrac, Relational Data Mining, Springer, 2001
MA-INF 4304  Lab Cognitive Robotics

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>every semester</td>
</tr>
</tbody>
</table>

Module coordinator
Prof. Dr. Sven Behnke

Lecturer(s)
Prof. Dr. Sven Behnke

Programme
M. Sc. Computer Science

Mode
Optional

Semester
2. or 3.

Learning goals: technical skills
Participants acquire practical experience and in-depth knowledge in the design and implementation of perception and control algorithms for complex robotic systems. In a small group, they analyze a problem, realize a state-of-the-art solution, and evaluate its performance.

Learning goals: soft skills
Self-competences (time management, goal-oriented work, ability to analyze problems and to find practical solutions), communication skills (Work together in small teams, oral and written presentation of solutions, critical examination of implementations)

Contents
Robot middleware (ROS), simultaneous localization and mapping (SLAM), 3D representations of objects and environments, object detection and recognition, person detection and tracking, action recognition, action planning and control, mobile manipulation, human-robot interaction.

Prerequisites
Recommended:
At least 1 of the following:
MA-INF 4113 – Cognitive Robotics
MA-INF 4114 – Robot Learning

Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lab</td>
<td>8</td>
<td>4</td>
<td>60 T / 210 S</td>
<td>9</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

Graded exams
Oral presentation, written report

Ungraded coursework (required for admission to the exam)

Literature
• B. Siciliano, O. Khatib (Eds.): Springer Handbook of Robotics, 2008.
• Selected research papers.
MA-INF 4306  Lab Development and Application of Data Mining and Learning Systems

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

Module coordinator: Prof. Dr. Stefan Wrobel

Lecturer(s): Prof. Dr. Stefan Wrobel

Programme: M. Sc. Computer Science

Mode: Optional

Semester: 3.

Learning goals: technical skills

Students will acquire in-depth knowledge in the construction and development of intelligent learning systems for machine learning and data mining. They learn how to work with existing state-of-the-art systems and apply them to application problems, usually extending them for the requirements of their particular task.

Learning goals: soft skills

Communicative skills (appropriate oral presentation and written documentation of project results), social skills (ability to work in teams), self-competences (time management, aiming at long-range goals under limited resources, ability to work under pressure, ability to accept/formulate criticism)

Contents

Data storage and process models of data analysis. Common open source frameworks for the construction of data analysis systems, specialized statistical packages. Pre-processing tools. Mathematical libraries for numerical computation. Search and optimization methods. User interfaces and visualization for analysis systems. Data analysis algorithms for embedded and distributed systems. Ubiquitous discovery systems.

Prerequisites

Recommended:

At least 1 of the following:

MA-INF 4111 – Intelligent Learning and Analysis Systems: Machine Learning
MA-INF 4112 – Intelligent Learning and Analysis Systems: Data Mining and Knowledge Discovery

Course meetings

Teaching format | Group size | h/week | Workload[h] | CP |
Lab            | 8          | 4       | 60 T / 210 S | 9   |

T = face-to-face teaching
S = independent study

Graded exams

Oral presentation, written report

Ungraded coursework (required for admission to the exam)

Forms of media

Computer Software, Documentation, Research Papers.

Literature

The relevant literature will be announced towards the end of the previous semester.
MA-INF 4308 Lab Vision Systems

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>every semester</td>
</tr>
</tbody>
</table>

**Module coordinator**
Prof. Dr. Sven Behnke

**Lecturer(s)**
Dr. Nils Goerke

**Programme**
M. Sc. Computer Science

**Mode**
Optional

**Semester**
3.

**Learning goals: technical skills**
Students will acquire knowledge of the design and implementation of parallel algorithms on GPUs. They will apply these techniques to accelerate standard machine learning algorithms for data-intensive computer vision tasks.

**Learning goals: soft skills**
Self-competences (time management, goal-oriented work, ability to analyze problems and to find practical solutions), communication skills (Work together in small teams, oral and written presentation of solutions, critical examination of implementations)

**Contents**
Basic matrix and vector computations with GPUs (CUDA). Classification algorithms, such as multi-layer perceptrons, support-vector machines, k-nearest neighbors, linear-discriminant analysis. Image preprocessing and data handling. Quantitative performance evaluation of learning algorithms for segmentation and categorization.

**Prerequisites**
Recommended:
At least 1 of the following:
MA-INF 4111 – Intelligent Learning and Analysis Systems: Machine Learning
MA-INF 4204 – Technical Neural Nets

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload [h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lab</td>
<td>8</td>
<td>4</td>
<td>60 T / 210 S</td>
<td>9</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

**Graded exams**
Oral presentation, written report

**Ungraded coursework (required for admission to the exam)**

**Literature**
MA-INF 4309 Lab Sensor Data Interpretation

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>at least every 2 years</td>
</tr>
</tbody>
</table>

Module coordinator: PD. Dr. Volker Steinhage
Lecturer(s): PD. Dr. Volker Steinhage

Programme: M. Sc. Computer Science
Mode: Optional
Semester: 2. or 3.

Learning goals: technical skills
Competence to implement algorithms for sensor data interpretation, efficient handling and testing, documentation.

Learning goals: soft skills
Efficient implementation of complex algorithms, abstract thinking, documentation of source code.

Contents
Varying selected up-to-date topics on sensor data interpretation

Prerequisites
Required:
MA-INF 2201 – Computer Vision

Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload</th>
<th>h</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lab</td>
<td>8</td>
<td>4</td>
<td>60 T / 210 S</td>
<td>9</td>
<td></td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

Graded exams
Oral presentation, written report

Ungraded coursework (required for admission to the exam)

Literature
Relevant literature will be announced at start of the lab.
MA-INF 4310  Lab Mobile Robots

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>at least every year</td>
</tr>
</tbody>
</table>

Module coordinator
Prof. Dr. Sven Behnke

Lecturer(s)
Prof. Dr. Sven Behnke, Dr. Nils Goerke

Programme
M. Sc. Computer Science

Mode
Optional

Semester
2. or 3.

Learning goals: technical skills
Participants acquire basic knowledge and practical experience in the design and implementation of control algorithms for simple structured robotic systems using real mobile robots. Fundamental paradigms for mobile robots will be identified and implemented in 2 person groups.

Learning goals: soft skills
Self-competences (time management, goal-oriented work, ability to analyze problems and to find practical solutions), communication skills (Work together in small teams, oral and written presentation of solutions, critical examination of implementations)

Contents
Robot middleware (e.g. ROS), robot simulation tools, basic capabilities for mobile robots: reactive control, SMPA architecture, navigation, path planning, localisation, simultaneous localization and mapping (SLAM), visual based object detection, learning robot control.

Prerequisites
Recommended:
At least 1 of the following:
- BA-INF 132 – Grundlagen der Robotik
- BA-INF 131 – Intelligente Sehsysteme
- MA-INF 1314 – Online Motion Planning
- MA-INF 2201 – Computer Vision
- MA-INF 4113 – Cognitive Robotics
- MA-INF 4114 – Robot Learning
- MA-INF 4203 – Autonomous Mobile Systems

Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lab</td>
<td>8</td>
<td>4</td>
<td>60 T / 210 S</td>
<td>9</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

Graded exams
Oral presentation, written report

Ungraded coursework (required for admission to the exam)

Forms of media
Robots simulation environments, robot control middleware, computer vision libraries, programming, demonstration of robot capabilities (real robotic systems), presentation and written report of approach and results.

Literature
- J. Buchli: Mobile Robots: Moving Intelligence, Published by Advanced Robotic Systems and Pro Literatur Verlag
- Additional State-of-the-art publications.
MA-INF 4312  Semantic Data Web Technologies

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>180 h</td>
<td>6 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

Module coordinator
Prof. Dr. Jens Lehmann

Lecturer(s)
Prof. Dr. Jens Lehmann, Dr. Christoph Lange, Dr. Maria Maleschkova

Programme
M. Sc. Computer Science

Mode
Optional

Semester
1.

Learning goals: technical skills

The goal of this lecture is to impart knowledge on the fundamentals, technologies and applications of the Semantic Web and information retrieval. As part of the lecture the basic concepts and standards for semantic technologies are explained.

Learning goals: soft skills

Contents

As part of the W3C Semantic Web initiative standards and technologies have been developed for machine-readable exchange of data, information and knowledge on the Web. These standards and technologies are increasingly being used in applications and have already led to a number of exciting projects (e.g. DBpedia, semantic wiki or commercial applications such as schema.org, OpenCalais, or Google’s Freebase). The module provides a theoretically grounded and practically oriented introduction to this area. The topics discussed within the lecture include:

- RDF syntax and data model
- RDF Schema and formal semantics of RDF (S)
- ontologies in OWL and formal semantics of OWL
- RDF databases, triple and knowledge stores, query languages
- Linked Data Web and Semantic Web applications
- Semantic text analysis and information retrieval systems

Prerequisites

none

Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td>2</td>
<td>30 T / 45 S</td>
<td>2.5</td>
<td></td>
</tr>
<tr>
<td>Exercises</td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
<td></td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

Graded exams

Written exam

Ungraded coursework (required for admission to the exam)

Successful exercise participation
### MA-INF 4313 Seminar Semantic Data Web Technologies

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>120 h</td>
<td>4 CP</td>
<td>1 semester</td>
<td>at least every year</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Module coordinator</th>
<th>Lecturer(s)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Prof. Dr. Jens Lehmann</td>
<td>Dr. Christoph Lange, Dr. Maria Maleshkova</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Programme</th>
<th>Mode</th>
<th>Semester</th>
</tr>
</thead>
<tbody>
<tr>
<td>M. Sc. Computer Science</td>
<td>Optional</td>
<td>2.</td>
</tr>
</tbody>
</table>

### Learning goals: technical skills
Through the seminar, students will learn to work with tools and technologies of the Semantic Web as well as assess their capabilities for given problems. They will gain the ability to understand new research results presented in original scientific papers.

### Learning goals: soft skills
Ability to present and to critically discuss technologies and research results in the framework of Semantic Web technologies.

### Contents
- technologies such as triple stores, link discovery frameworks, NLP pipelines.
- recent conference and journal papers

### Prerequisites
none

### Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Seminar</td>
<td>10</td>
<td>2</td>
<td>30 T / 90 S</td>
<td>4</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

### Graded exams
Oral presentation, written report

### Ungraded coursework (required for admission to the exam)
MA-INF 4314  Lab Semantic Data Web Technologies

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

**Module coordinator**

Prof. Dr. Jens Lehmann

**Lecturer(s)**

Prof. Dr. Jens Lehmann, Dr. Maria Maleschkova

**Programme**

M. Sc. Computer Science

**Mode**

Optional

**Semester**

2.

**Learning goals: technical skills**

The students will carry out a practical task (project) in the context of Semantic Web technologies, including test and documentation of the implemented software/system.

**Learning goals: soft skills**

Ability to properly present and defend design decisions, to prepare readable documentation of software; skills in constructively collaborating with others in small teams over a longer period of time; ability to classify own results with regard to the state-of-the-art

**Contents**

**Prerequisites**

none

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lab</td>
<td>8</td>
<td>4</td>
<td>60 T / 210 S</td>
<td>9</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

**Graded exams**

Oral presentation, written report

**Ungraded coursework (required for admission to the exam)**
MA-INF 4316  Graph Representation Learning

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>180 h</td>
<td>6 CP</td>
<td>1 semester</td>
<td>at least every 2 years</td>
</tr>
</tbody>
</table>

**Module coordinator**
Dr. Pascal Welke

**Lecturer(s)**
Dr. Pascal Welke

**Programme**
M. Sc. Computer Science

**Mode**
Optional

**Semester**
1.

**Learning goals: technical skills**
- Deep understanding of the trade-off between expressiveness of graph representation and computational complexity, as well as practical runtime of algorithms in the context of machine learning applications.
- Ability to implement, practically apply, and theoretically analyze graph representation, graph kernels, and graph mining algorithms.

**Learning goals: soft skills**
- Social, methodological, and analytical competences via communication, own development, and presentation of problem formulations, algorithms, and solutions.
- Learning to solve project tasks in a group.
- Learning to evaluate the trade-offs and limitations of existing methods.

**Contents**
We will discuss general approaches for machine learning (ML) on graph structured data. In particular, computational methods for graph representation learning such as graph neural networks (GNNs), graph kernels, as well as graph mining techniques will be discussed, analyzed, and applied. Regarding GNNs and graph kernels, we will discuss the expressive power and how these concepts are related, as well as several specific examples. In the area of graph mining, we will likely investigate fast (approximate) algorithms to count small patterns, such as triangles, or trees.

If time permits, we might venture into the realm of ranking on large-scale graphs, with applications such as recommender systems. The exercises will focus on practical implementations and the application of these methods to real world examples.

**Prerequisites**
**Recommended:**
Helpful: one or more of the following
- MA-INF 4111 – Intelligent Learning and Analysis Systems: Machine Learning
- MA-INF 4112 – Intelligent Learning and Analysis Systems: Data Mining and Knowledge Discovery
- MA-INF 4212 – Data Science and Big Data
- MA-INF 1105 - Algorithms for Data Analysis
- MA-INF 1102 - Combinatorial Optimization

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
<th>T = face-to-face teaching</th>
<th>S = independent study</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td>2</td>
<td>30 T / 45 S</td>
<td>2.5</td>
<td>3</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Exercises</td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
<td></td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

**Graded exams**
Oral exam or written exam

**Ungraded coursework (required for admission to the exam)**
Successful exercise participation

**Forms of media**
- Lecture slides
- Jupyter notebooks
Literature

• William L. Hamilton: Graph Representation Learning, Synthesis Lectures on Artificial Intelligence and Machine Learning, Morgan and Claypool.
**MA-INF 4318 Seminar Representation Learning for Big Data Analytics**

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>120 h</td>
<td>4 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

**Module coordinator**
Prof. Dr. Emmanuel Müller

**Lecturer(s)**
Prof. Dr. Emmanuel Müller

**Programme**
M. Sc. Computer Science

**Mode**
Optional

**Semester**
2. or 3.

**Learning goals: technical skills**
Ability to understand new research results presented in original scientific papers.

**Learning goals: soft skills**
Ability to present and to critically discuss these results in the framework of the corresponding area.

**Contents**
Smart representations (such as embeddings, kernels, and dimensionality reduction methods) are useful models that allow the abstraction of data within a well-defined mathematical formalism. The representations we aim at are conceptual abstractions of real world phenomena (such as social interactions, chemical reactions and biological processes) into the world of statistics and discrete mathematics in such a way that the powerful tools developed in those areas are available for complex analyses in a simple and elegant manner.

The focus will be the understanding and comparison of smart representations and their explicit/implicit data transformation models. We will study limitations and advantages of different techniques, and how the data representation changes the problem setup, reduces complexity, introduces robustness, or other valuable properties for big data analytics.

**Prerequisites**
Recommended:
Open-minded for new problem settings, Programming in different languages (C++, Python, Java), Critical approach to existing solutions, Research curiosity

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Seminar</td>
<td>10</td>
<td>2</td>
<td>30 T / 90 S</td>
<td>4</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

**Graded exams**
Oral presentation, written report

**Ungraded coursework (required for admission to the exam)**

**Literature**

MA-INF 4322  Lab Machine Learning on Encrypted Data

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

**Module coordinator**
Dr. Michael Nüsken

**Lecturer(s)**
Dr. Michael Nüsken

**Programme**
M. Sc. Computer Science

**Frequency**
every year

**Learning goals: technical skills**
The students will carry out a practical task (project) in the context of Cryptography, including test and documentation of the implemented software/system.

**Learning goals: soft skills**
Ability to properly present and defend design decisions, to prepare readable documentation of software; skills in constructively collaborating with others in small teams over a longer period of time; ability to classify one's own results into the state-of-the-art of the resp. area

**Contents**
With the rise of more and more mechanisms and installations of data science methodology to automatically analyze large amounts of possibly privacy infringing data we have to carefully understand how to protect our data. Also more and more fake data shows up and we have to find ways to distinguish faked from trustable data. At the same time we want to allow insightful research and life-easing analyzes to be possible. This seeming contradiction has lead to various efforts for unifying both: protecting data and allowing analyzes, at least to some extent and possibly under some restrictions. See Munn et al. (2019) for a review on challenges and options.

The target of the lab is to understand how computations on encrypted data may work in one particular application that we are choosing together. Ideally, we can come up with a novel solution for performing an unconsidered algorithm. We study the tasks and tools, select algorithms, find a protocol, prototype an implementation, perform a security analysis, present an evaluation, ...

**Prerequisites**
Recommended:
Basic knowledge in cryptography is highly recommended, eg. by MA-INF 1103 - Cryptography, MA-INF 1223 - PETs, MA-INF 1209 - Seminar Advanced Topics in Cryptography.

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td>4</td>
<td>60 T / 105 S</td>
<td>5.5</td>
<td></td>
</tr>
<tr>
<td>Exercises</td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
<td></td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

**Graded exams**
Written exam

**Ungraded coursework (required for admission to the exam)**
Successful exercise participation
MA-INF 4324  Seminar Advanced Topics in Data Science

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>120 h</td>
<td>4 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

Module coordinator
Prof. Dr. Elena Demidova

Lecturer(s)
Prof. Dr. Elena Demidova

Programme
M. Sc. Computer Science

Mode
Optional

Semester
2. or 3.

Learning goals: technical skills
This module concentrates on specialized topics in data science. The students obtain skills in the independent, in-depth study of state-of-the-art scientific literature on specific topics, discussion with their peers and presentation to the scientific audience.

Learning goals: soft skills
- Communication skills: oral and written presentation of scientific content.
- Self-competences: the ability to analyze problems, time management, creativity.

Contents
Statistical and machine learning-based methods of data analytics, including typical steps of the data science process: data generation, integration, cleaning, exploration, modelling and evaluation. Specialized data representation and analytics methods for selected data types and applications in specific domains.

Prerequisites
Recommended:
BA-INF 150 - Einführung in die Data Science

Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[hr]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Seminar</td>
<td>10</td>
<td>2</td>
<td>30 T / 90 S</td>
<td>4</td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

Graded exams
Oral presentation, written report

Ungraded coursework (required for admission to the exam)
None

Literature
Relevant literature will be announced at the beginning of the seminar
MA-INF 4325  Lab Data Science in Practice

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

**Module coordinator**
- Prof. Dr. Elena Demidova

**Lecturer(s)**
- Prof. Dr. Elena Demidova

**Programme**
- M. Sc. Computer Science

**Mode**
- Optional

**Semester**
- 2. or 3.

**Learning goals: technical skills**
This module concentrates on practical experience in data analytics. Participants acquire basic knowledge and practical experience in the design and implementation of data science workflows for specific data types and applications.

**Learning goals: soft skills**
- Communication skills: the ability to work in teams.
- Self-competences: the ability to analyse problems and find practical solutions. Time management, creativity, presentation of results.

**Contents**
Practical application of statistical and machine learning-based methods to solve data analytics problems on real-world datasets and evaluate proposed solutions.

**Prerequisites**
**Recommended:**
- BA-INF 150 - Einführung in die Data Science
- MA-INF 4230 - Advanced Methods of Information Retrieval

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lab</td>
<td>8</td>
<td>4</td>
<td>60 T / 210 S</td>
<td>9</td>
</tr>
</tbody>
</table>

**Graded exams**
- Oral presentation, written report

**Ungraded coursework (required for admission to the exam)**
- None
MA-INF 4326  Explainable AI and Applications

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>180 h</td>
<td>6 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

**Module coordinator**  
Dr. Tiansi Dong  
**Lecturer(s)**  
Dr. Tiansi Dong

**Programme**  
M. Sc. Computer Science  
**Mode**  
Optional  
**Semester**  
3.

**Learning goals: technical skills**
- Know the dual-model functioning of the human mind, and two main AI paradigms
- Develop white-box neural AI systems
- Understand the problems and limitations of Blackbox Deep-Learning systems, and Know the state-of-the-art Methods for Interpreting Deep-Learning systems (XAI)

**Learning goals: soft skills**
- Know System 1 and 2 of the mind, prons and cons of symbolic AI and connectionist AI
- Develop neural-geometric systems that have both good features of symbolic AI and connectionist AI
- Know the limitation of famous Deep-Learning systems, such as GPT3, self-driving. Know standard methods to explore the explainability of Deep-Learning systems

**Contents**
1. Introduction: fates of large Deep-Learning systems, e.g. Watson, GPT, self-driving cars
2. Dual-system theories (System 1 and 2), nine laws of cognition, criteria of semantic models
3. The target and the state-of-art methods of XAI
4. Neural-symbolic AI
5. Cognitive maps, Collages, Mental Spatial Representation, Events
6. Qualitative Spatial Representation and Reasoning
7. Rotating Sphere Embedding: A New Wheel for Neural-Symbolic Unification
8. Neural Syllogistic Reasoning
9. Recognizing Variable Environments
10. Humor Understanding
11. Rotating Spheres as building-block semantic components for Language, Vision, and Action

**Prerequisites**
none

**Course meetings**
- **Teaching format**
  - **Lecture**: 2 h/week, 30 T / 45 S, 2.5 CP  
  - **Exercises**: 2 h/week, 30 T / 75 S, 3.5 CP

  *T = face-to-face teaching  
  *S = independent study

**Graded exams**
Written exam

**Ungraded coursework (required for admission to the exam)**
The completion of regularly provided exercise sheets. The work can be done in groups of up to four students. A total of 50% of the points must be achieved.
Literature

# MA-INF 4327 Lab Biomedical Data Science

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>270 h</td>
<td>9 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

**Module coordinator**  
Prof. Dr. Holger Fröhlich  
**Lecturer(s)**  
Prof. Dr. Holger Fröhlich

**Programme**  
M. Sc. Computer Science  
**Mode**  
Optional  
**Semester**  
3.

**Learning goals: technical skills**  
The students will carry out a practical task (project) in the context of biomedical data science, including test and documentation of the implemented software/system.

**Learning goals: soft skills**  
Ability to properly present and defend design decisions, to prepare readable documentation of software; skills in constructively collaborating with others in small teams over a longer period of time; ability to classify ones own results into the state-of-the-art of the resp. area

**Contents**  
Varying selected topics close to current research in the area of biomedical data science.

**Prerequisites**  
none

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lab</td>
<td>8</td>
<td>4</td>
<td>60 T / 210 S</td>
<td>9</td>
</tr>
</tbody>
</table>

T = face-to-face teaching  
S = independent study

**Graded exams**

Oral presentation, written report

**Ungraded coursework (required for admission to the exam)**
MA-INF 4328  Spatio-Temporal Data Analytics

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>180 h</td>
<td>6 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

Module coordinator
Prof. Dr. Elena Demidova  
Lecturer(s)
Prof. Dr. Elena Demidova

Programme
M. Sc. Computer Science  
Mode
Optional  
Semester
2. or 3.

Learning goals: technical skills
This module introduces the students to the advanced methods, data structures, and data analytics algorithms for spatio-temporal data. At the end of the module, the students will be capable of choosing appropriate data representations, data structures and algorithms for specific applications and correctly applying relevant statistical and machine learning-based data analytics procedures.

Learning goals: soft skills
Communication skills: oral and written presentation and discussion of solutions. Self-competences: the ability to analyze and solve problems.

Contents
The module topics include data structures, data representation and analysis methods, and algorithms that enable analyzing spatio-temporal data and building predictive models effectively and effectively. Furthermore, we will study the corresponding evaluation techniques and novel applications.

Prerequisites
none

Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Lecture</td>
<td></td>
<td>2</td>
<td>30 T / 45 S</td>
<td>2.5</td>
</tr>
<tr>
<td>Exercises</td>
<td></td>
<td>2</td>
<td>30 T / 75 S</td>
<td>3.5</td>
</tr>
</tbody>
</table>

T = face-to-face teaching  
S = independent study

Graded exams
Written exam

Ungraded coursework (required for admission to the exam)
The completion of regularly provided exercise sheets. The work can be done in groups of up to three, four or five students, depending on the total number of students taking the course. A total of 50% of the points must be achieved. For 80% of the exercise sheets, 40% of the points must be achieved for each sheet. Each student must present a solution to an exercise in the exercise sessions once.
MA-INF 4329  Seminar Biological Intelligence

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>120 h</td>
<td>4 CP</td>
<td>1 semester</td>
<td>every year</td>
</tr>
</tbody>
</table>

**Module coordinator**
Prof. Dr. Dr. Dominik Bach

**Lecturer(s)**
Prof. Dr. Dr. Dominik Bach

**Programme**
M. Sc. Computer Science

**Mode**
Optional

**Semester**
2. or 3.

**Learning goals: technical skills**
Ability to understand new research results presented in original scientific papers.

**Learning goals: soft skills**
Ability to present and to critically discuss these results in the framework of the corresponding area.

**Contents**
Current conference and journal papers.

**Prerequisites**
none

**Course meetings**

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload</th>
<th>h</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Seminar</td>
<td>10</td>
<td>2</td>
<td>30 T / 90 S</td>
<td>4</td>
<td></td>
</tr>
</tbody>
</table>

T = face-to-face teaching
S = independent study

**Graded exams**
Oral presentation, written report

**Ungraded coursework (required for admission to the exam)**
## Master Thesis

<table>
<thead>
<tr>
<th>Course</th>
<th>Credits</th>
<th>Title</th>
<th>Page</th>
</tr>
</thead>
<tbody>
<tr>
<td>MA-INF 0401</td>
<td>30 CP</td>
<td>Master Thesis</td>
<td>154</td>
</tr>
<tr>
<td>MA-INF 0402</td>
<td>2 CP</td>
<td>Master Seminar</td>
<td>155</td>
</tr>
</tbody>
</table>
MA-INF 0401  Master Thesis

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>900 h</td>
<td>30 CP</td>
<td>1 semester</td>
<td>every semester</td>
</tr>
</tbody>
</table>

Module coordinator

Lecturer(s)

All lecturers of computer science

Programme

M. Sc. Computer Science

Mode

Compulsory

Semester

4.

Learning goals: technical skills

Ability to solve a well-defined, significant research problem under supervision, but in principle independently

Learning goals: soft skills

Ability to write a scientific documentation of considerable length according to established scientific principles of form and style, in particular reflecting solid knowledge about the state-of-the-art in the field

Contents

Topics of the thesis may be chosen from any of the areas of computer science represented in the curriculum

Prerequisites

none

Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Independent preparation of a scientific thesis with individual coaching</td>
<td>0</td>
<td>900 S</td>
<td>30</td>
<td></td>
</tr>
</tbody>
</table>

T = face-to-face teaching

S = independent study

Graded exams

Master Thesis

Ungraded coursework (required for admission to the exam)

Literature

Individual bibliographic research required for identifying relevant literature (depending on the topic of the thesis)
MA-INF 0402  Master Seminar

<table>
<thead>
<tr>
<th>Workload</th>
<th>Credit points</th>
<th>Duration</th>
<th>Frequency</th>
</tr>
</thead>
<tbody>
<tr>
<td>60 h</td>
<td>2 CP</td>
<td>1 semester</td>
<td>every semester</td>
</tr>
</tbody>
</table>

Module coordinator

Lecturer(s)
All lecturers of computer science

Programme

M. Sc. Computer Science

Mode
Compulsory

Semester
4.

Learning goals: technical skills
Ability to document and defend the results of the thesis work in a scientifically appropriate style, taking into consideration the state-of-the-art in research in the resp. area

Learning goals: soft skills

Contents
Topic, scientific context, and results of the master thesis

Prerequisites
none

Course meetings

<table>
<thead>
<tr>
<th>Teaching format</th>
<th>Group size</th>
<th>h/week</th>
<th>Workload[h]</th>
<th>CP</th>
</tr>
</thead>
<tbody>
<tr>
<td>Seminar</td>
<td></td>
<td>2</td>
<td>30 T / 30 S</td>
<td>2</td>
</tr>
</tbody>
</table>

T = face-to-face teaching  
S = independent study

Graded exams

Oral presentation of final results

Ungraded coursework (required for admission to the exam)

Literature

Individual bibliographic research required for identifying relevant literature (depending on the topic of the thesis)