University Hildesheim

Faculty 4 Mathematics, Natural Sciences, Economics and Computer Science



Data Analytics International Master

Course Catalogue (Fundamental Track)

Version PO 2023 as of 29. September 2023 last editorial changes: October 11, 2023

Compulsory Modules

Module	Type/HPW	\mathbf{CPs}	P.
Machine Learning	2 HPW lecture, 2 HPW tutorial	6	5
Modern Optimization Techniques	2 HPW lecture, 2 HPW tutorial	6	7
Advanced Machine Learning	2 HPW lecture, 2 HPW tutorial	6	9
Big Data Analytics	2 HPW lecture, 2 HPW tutorial	6	11
Planning and Optimal Control	2 HPW lecture, 2 HPW tutorial	6	12
Lab Course Programming Machine Learning	4 HPW lab course	6	14
Lab Course Distributed Data Analytics	4 HPW lab course	6	15
Data and Privacy Protection	2 HPW lecture	3	16
Seminar Data Science II	2 HPW Seminar	4	17
Seminar Data Analytics I	2 HPW seminar	4	18
Seminar Data Analytics II	2 HPW seminar	4	19
Student Research Project	4 HPW project	15	20

Elective Modules

Methodological Specialization

Module	$\mathbf{Type}/\mathbf{HPW}$	CPs	P.
Advanced Case-Based Reasoning	2 HPW lecture, 2 HPW tutorial	6	21
Advanced Computer Vision	2 HPW lecture, 2 HPW tutorial	6	23
Bayesian Networks	2 HPW lecture, 2 HPW tutorial	6	24
Business Analytics	2 HPW lecture, 2 HPW tutorial	6	26
Computer Vision	2 HPW lecture, 2 HPW tutorial	6	28
Constraint Logic Programming	2 HPW lecture and tutorial	3	29
Deep Learning	2 HPW lecture, 2 HPW tutorial	6	30
Deep Learning 2	2 HPW lecture, 2 HPW tutorial	6	31
Deep Learning Masterclass	4 HPW lecture, 4 HPW tutorial	6	32
Graph Analytics	2 HPW lecture, 2 HPW tutorial	6	33
Lab Course Deep Learning	4 HPW lab course	6	34
Machine Learning for IT Security	2 HPW lecture, 2 HPW tutorial	6	35
Survey Sampling	2 HPW lecture, 2 HPW tutorial	6	37
Time Series Analysis	2 HPW lecture, 2 HPW tutorial	6	38

Application

Computer Science/Software Engineering

Module	Type/HPW	CPs	P.
Software Engineering	2 HPW lecture, 2 HPW tutorial	6	39
Model-based Software Engineering	2 HPW lecture, 2 HPW tutorial	6	41
Advanced Web Development	2 HPW lecture, 2 HPW lab	6	42
	course		
Processes and Management of Software Engineering	2 HPW lecture, 2 HPW tutorial	6	43
Software Product Line Engineering	2 HPW lecture, 2 HPW tutorial	6	44
Software Testing	2 HPW lecture (with integrated	3	46
	tutorial)		
Analysis of Software Systems	2 HPW Lecture with Tutorium	3	47
Software Verification	2 HPW lecture and tutorial	3	48
Software Architectures	2 HPW lecture, 2 HPW tutorial	6	49
Requirements Engineering for Quality	2 HPW lecture, 2 HPW tutorial	6	51
Constraint Logic Programming	2 HPW lecture and tutorial	3	29
SAT Solving	2 HPW Seminar	4	52
Efficient Algorithms			53
SE Tools Lab	4 HPW lab course	6	54
Master-Seminar Software Engineering	2 HPW seminar	4	55

Business Administration and Information Systems

Module	Type/HPW	CPs	P.
Advanced Marketing	3 HPW lecture	4	56
Product development and technologies for navigation	2 HPW lecture	3	57
and driver assistance systems			
Project Management and Scheduling	2 HPW lecture, 2 HPW tutorial	6	59
Seminar Business Studies	2 HPW seminar	4	61
Business Intelligence and Data Warehousing	2 HPW lecture, 2 HPW tutorial	6	62
Data Warehousing in Practice	2 HPW lecture, 2 HPW tutorial	6	63

Information Retrieval and Information Sciences

Module	$\mathbf{Type}/\mathbf{HPW}$	CPs	
Introduction Information Retrieval (IR)	2 HPW lecture	4	65
Introduction Natural Language Processing	3 HPW lecture	4	66
Multilingual Information Systems	2 HPW lecture	4	67
Seminar Multilingiual Information Retrieval	2 HPW seminar	4	68
Project Multilingual Information Systems	4 HPW project	6	69
Lab Course Information Retrieval (IR)	2 HPW lab course	4	70

Natural Language Processing

Module	Type/HPW	\mathbf{CPs}	P.
Natural Language Processing	2 HPW lecture, 2 HPW tutorial	6	71
Language Modelling	2 HPW lecture 2 HPW project	6	73
Natural Language Processing 2	2 HPW lecture	4	75
Seminar Computer Linguistic Resources	2 HPW seminar	4	77
Seminar Computer Linguistic Processes	2 HPW seminar	4	78
Project Computer Linguistic Resources	4 HPW project	6	79
Project Computer Linguistic Processes	4 HPW project	6	80
Lab Course Natural Language Processing	2 HPW lab course	4	81

Psychology

Module	Type/HPW	\mathbf{CPs}	P.
Cognitive Psychology	2 HPW lecture, 2 HPW tutorial	6	83
Brain and Neural Science	2 HPW lecture, 2 HPW tutorial	6	84

Environmental Sciences

Module	$\mathrm{Type}/\mathrm{HPW}$	\mathbf{CPs}	P.
Geographic Information Systems	2 HPW lecture, 4 HPW tutorial	6	87

Soft Skills

Module	${ m Type/HPW}$	CPs	P.
English 1	2 HPW lecture	3	88
English 2	2 HPW lecture	3	89
German 1	2 HPW lecture	3	90
German 2	2 HPW lecture	3	91
International Start-up School	2 HPW lecture	3	92

Master Thesis

Module	Type/HPW	\mathbf{CPs}	P.
Master Thesis	Master Thesis	30	93

Compulsory Modules

Module: Machine Learning

Responsible	Prof. Dr. Lars Schmidt-Thieme
Lecturer	none
Type	2 HPW lecture, 2 HPW tutorial
Credit Points	6 CPs
Workload	presence: 42 hours; self-study: 108 hours
Learning goals/ Competencies	After the completion of this module, the students should be able to map practical tasks to their respective theoretical problem. They should have developed a deeper understanding in the field of machine learning. They should be able to recognize the different types of machine learning problems as well as understand, implement and apply different machine learning techniques. The students should be capable of adapting those techniques to specific applications. In addition, they should be in a position to understand and elaborate further procedures based on the literature.
Content	 The lecture gives an overview of machine learning. It focuses on: fundamental machine learning problems: different machine learning problems are described and shown in examples. classification: basic models for decision and classification tasks are treated (logistic regression, nearest neighbor, decision trees, neuronal networks, support-vector machine, simple bayesian networks). cluster-analysis and dimensionality reduction: models for non-supervised classification are treated (hierarchical clustering, k-means, graph partitioning). application of machine learning models for problems in informatics
Submodules	SM 1: Advanced Machine Learning, Lecture Type: 2 HPW Lecture (3 CPs) Lecturer: Prof. Dr. Dr. Lars Schmidt-Thieme SM 2: Advanced Machine Learning, Tutorium Type: 2 HPW Tutorium (3 CPs) Lecturer: Prof. Dr. Dr. Lars Schmidt-Thieme and members of the study group

Literature	 Kevin Murphy: Machine Learning: a Probabilistic Perspective. MIT Press, 2012. Richard O. Duda, Peter E. Hart, David G. Stork: Pattern Classification. Springer, 2001. Trevor Hastie, Robert Tibshirani, Jerome Friedman: The Elements of Statistical Learning. Springer, 2001. Tom Mitchell: Machine Learning. McGraw-Hill, 1997.
Requirements	The module Machine Learning is compulsory for all students but those having an equivalent module already in their Bachelor's.
Exam	written exam (120 min) or an oral exam (30 min)
Recommended Term	MSc 1
Turn	every winter term
Duration	1 Semester
Use	• MSc AInf Variant: Data Analytics (fundamental track) – Compulsory Modules
Language	English

Module: Modern Optimization Techniques

Responsible	Prof. Dr. Lars Schmidt-Thieme
Type	2 HPW lecture, 2 HPW tutorial
Credit Points	6 CPs
Workload	presence: 42 hours; self-study: 108 hours
Learning goals/ Competencies	The students should have developed a deeper understanding in the field of Optimization. They learn to implement and apply different optimization techniques and should be able to adapt these techniques to specific applications. They should be able to map practical tasks to their respective theoretical problem. Students are able to recognize different types of optimization problems and should be able to understand and elaborate further procedure based on the literature.
Content	The Lecture will discuss Optimization techniques on which modern Data Analytics approaches are based. The topics discussed will be:
	 Optimization Problems: the different types of Optimization problems will be described both formally and with examples Unconstrained and Equality Constrained Convex Optimization: The main convex optimization methods (Stochastic Gradient Descent, Newton Methods, and Coordinate Descent) Interior Point Methods: Methods for solving inequality constrained problems by solving a sequence of unconstrained, or equality constrained, problems. Modern Optimization methods: Extensions and improvements of classical optimization methods: Quasi-Newton, Conjugate Gradient, Bundle methods and Cutting-plane algorithms
Submodules	SM 1: Modern Optimization Techniques, lecture Type: 2 HPW lecture (3 ECTS) Lecturer: Prof. Dr. Dr. Lars Schmidt-Thieme SM 2: Modern Optimization Techniques, tutorial Type: 2 HPW tutorial (3 ECTS) Lecturer: Prof. Dr. Dr. Lars Schmidt-Thieme and members of the study group
Literature	· · ·
	 Stephen Boyd and Lieven Vandenberghe. Convex Optimization. Cambridge Univ Press, 2004. Suvrit Sra, Sebastian Nowozin and Stephen J. Wright. Optimization for Machine Learning. MIT Press, 2011. Igor Griva. Linear and nonlinear optimization. Society for Industrial and Applied Mathematics, 2009.
Requirements	none
Exam	written exam (120 min) or an oral exam (30 min)
Recommended Term	MSc 1
Turn	every winter term
Duration	1 Semester

Use	• MSc AInf Variant: Data Analytics (fundamental track) – Compulsory Modules
Language	English

Module: Advanced Machine Learning

Responsible	Prof. Dr. Lars Schmidt-Thieme
Lecturer	none
Type	2 HPW lecture, 2 HPW tutorial
Credit Points	6 CPs
Workload	presence: 42 hours; self-study: 108 hours
Learning goals/ Competencies	After the completion of this module, the students should be able to map practical tasks to their respective theoretical problem. They should have developed a deeper understanding in the field of machine learning. They should be able to recognize the different types of machine learning problems as well as understand, implement and apply different machine learning techniques. The students should be capable of adapting those techniques to specific applications. In addition, they should be in a position to understand and elaborate further procedures based on the literature.
Content	 The lecture gives an overview of machine learning. It focusses on: fundamental machine learning problems: different machine learning problems are described and shown in examples. classification: basic models for decsion and classification tasks are treated (logistic regression, nearest neighbor, decision trees, neuronal networks, support-vector machine, simple bayesian networks). cluster-analysis and dimensionality reduction: models for non-supervised classification are treated (hierarchical clustering, k-means, graph partitioning). methods for learning hyperparameters structured prediction
Submodules	SM 1: Advanced Machine Learning, Lecture Type: 2 HPW Lecture (3 CPs) Lecturer: Prof. Dr. Dr. Lars Schmidt-Thieme SM 2: Advanced Machine Learning, Tutorium Type: 2 HPW Tutorium (3 CPs) Lecturer: Prof. Dr. Dr. Lars Schmidt-Thieme and members of the study group
Literature	v U 1
	 Kevin Murphy: Machine Learning: a Probabilistic Perspective. MIT Press, 2012. Richard O. Duda, Peter E. Hart, David G. Stork: Pattern Classification. Springer, 2001. Trevor Hastie, Robert Tibshirani, Jerome Friedman: The Elements of Statistical Learning. Springer, 2001. Tom Mitchell: Machine Learning. McGraw-Hill, 1997.
Requirements	The lecture 'Machine Learning' is recommended.
Exam	written exam (120 min) or an oral exam (30 min)
Recommended Term	MSc 1
Turn	every summer term
Duration	1 Semester

Use	MSc AInf Variant: Data Analytics (fundamental track) – Compulsory Modules
Language	English

Module: Big Data Analytics

Responsible	Prof. Dr. Lars Schmidt-Thieme
Type	2 HPW lecture, 2 HPW tutorial
Credit Points	6 CPs
Workload	presence: 42 hours; self-study: 108 hours
Learning goals/ Compe-	The Students should have developed an extended understanding
tencies	in the field of Big Data Analytics and be able to map practical
	tasks to their respective theoretical problem. They should be able
	to use the learned methods for more complex problems and be able
	to recognize differences in the problems. In addition, they should
	be in a position to understand and elaborate further procedures
	based on the literature.
Content	The course will cover the following topic areas:
	1. Large Scale Distributed File Systems and Data Storage frame- works
	2. Computational models for large scale data: (e.g. MapReduce and GraphLab)
	3. Data Stream Analysis
	4. Statistical learning techniques for Large Scale Data: For example Large Scale Recommender Systems and Link Analysis
Submodules	SM 1: Big Data Analytics, lecture Type: 2 HPW lecture (3 CPs) Lecturer: Prof. Dr. Lars Schmidt-Thieme SM 2: Big Data Analytics, tutorial Type: 2 HPW tutorial (3 CPs) Lecturer: Prof. Dr. Dr. Lars Schmidt-Thieme and members of the study group
Literature	
	 Anand Rajaraman, Jure Leskovec, and Jeffrey Ullman: Mining of massive datasets Yucheng Low, Joseph Gonzalez, Aapo Kyrola, Danny Bickson, Carlos Guestrinand Joseph M. Hellerstein: Distributed GraphLab: A Framework for Machine Learning and Data Mining in the Cloud PVLDB. 2012
Requirements	none
Exam	written exam (120 min) or an oral exam (30 min)
Recommended Term	MSc 2
Turn	every 4th Semester but not regularly.
Duration	1 Semester
Use	• MSc AInf Variant: Data Analytics (fundamental track) – Compulsory Modules
Language	English

Module: Planning and Optimal Control

Responsible	Prof. Dr. Lars Schmidt-Thieme
Type	2 HPW lecture, 2 HPW tutorial
Credit Points	6 CPs
Workload	presence: 42 hours; self-study: 108 hours
Learning goals/ Competencies	After the completion of this module, the students should be able to
tencies	map practical tasks to their respective theoretical problem. They should have developed a deeper understanding in the field of Plan-
	ning and Optimal Control. They should be able to recognize the
	different types of planning and control problems as well as under-
	stand, implement and apply different techniques. The students
	should be capable of adapting those techniques to specific applica-
	tions. In addition, they should be in a position to understand and
	elaborate further procedures based on the literature.
Content	The lecture will discuss main topics from Planning and optimal control theory. The topics discussed will be:
	1. Discrete and Heuristic Search
	2. Motion Planning
	3. Dealing with dynamics and Stochastic Optimal Control
	4. Reinforcement Learning
Submodules	SM 1: Planning and Optimal Control, lecture Type: 2 HPW lecture (3 ECTS) Lecturer: Prof. Dr. Lars Schmidt-Thieme SM
	2: Planning and Optimal Control, tutorial Type: 2 HPW tutorial
	(3 ECTS) Lecturer: Prof. Dr. Dr. Lars Schmidt-Thieme and
	members of the study group
Literature	
	• H. Geffner, B. Bonet: A Concise Introduction to Models and Methods for Automated Planning, Morgan and Claypool, 2013.
	• D. Nau, M. Ghallab, P. Traverso: Automated Planning: Theory and Practice, Morgan Kaufmann, 2004.
	• H. Choset, K. M. Lynch, S. Hutchinson, G. Kantor, W. Bur-
	gard, L. E. Kavraki and S. Thrun. Principles of Robot Mo-
	tion: Theory, Algorithms, and Implementations; MIT Press,
	Boston, 2005.
	• Steve LaValle. <i>Planning Algorithms</i> ; Cambridge University
	Press, 2006 (Available Online). • Dimitri P. Bertsekas. Dynamic Programming and Optimal
	Control, Athena Scientific, 3rd ed. Vols. I and II, 2007.
	• Richard S. Sutton and Andrew G. Barto. Reinforcement
	Learning: An Introduction. MIT Press, Cambridge, MA,
	1998.
Requirements	none
Exam	written exam (120 min) or an oral exam (30 min)
Recommended Term	MSc 3

Turn	every winter term
Duration	1 Semester
Use	MSc AInf Variant: Data Analytics (fundamental track) – Compulsory Modules
Language	English

Module: Lab Course Programming Machine Learning

Responsible	Prof. Dr. Lars Schmidt-Thieme
Type	4 HPW lab course
Credit Points	6 CPs
Workload	presence: 42 hours; self-study: 108 hours
Learning goals/ Compe-	The praktikum allows students to gain practical knowledge and
tencies	capabilities in the area of Machine Learning.
Content	This implementation-oriented course offers hands-on experience with current algorithms and approaches in Machine Learning and Artificial Intelligence, and their application to real-world learning and decision-making tasks. Praktikum will also cover empirical methods for comparing learning algorithms, for understanding and explaining their differences, for analyzing the conditions in which a method is more suitable than others. List of Methods: 1. Linear models of prediction (Linear Regression, Logistic Regression) 2. Generative learning algorithms (Discriminant Analysis, Naïve Bayes) 3. Classification trees (Decision Trees) 4. k - Nearest Neighbor 5. Clustering (k-Means) 6. Dimensionality Reduction (Principal Component Analysis) 7. Support Vector Machines 8. Matrix Factorization for Recommender Systems
Submodules	none
Literature	
	 Brett Lantz: Machine Learning with R, Packt Publishing, 2013. Drew Conway, John Myles White: Machine Learning for Hackers, O'Reilly, 2012.
Requirements	none
Exam	colloquium and written summary
Recommended Term	MSc 1
Turn	every winter term
Duration	1 Semester
Use	• MSc AInf Variant: Data Analytics (fundamental track) – Compulsory Modules
Language	English

Module: Lab Course Distributed Data Analytics

Responsible	Prof. Dr. Lars Schmidt-Thieme
Type	4 HPW lab course
Credit Points	6 CPs
Workload	presence: 42 hours; self-study: 108 hours
Learning goals/ Compe-	The praktikum allows students to gain practical knowledge and
tencies	capabilities in the area of Machine Learning. Praktikum will also
	cover empirical methods for comparing learning algorithms, for
	understanding and explaining their differences, for analyzing the
	conditions in which a method is more suitable than others.
Content	Practical knowledge of methods and technologies for distributed
	computing in data analysis:
	1. working with a scheduler in a Computer Cluster (e.g. Sun
	Grid Engine)
	2. working with a distributed data system to manage big data 3. working with NoSQL-data-bases for loose structured data
	4. Large Scale distributed file systems and data storage frame-
	works
	5. Computational models for large scale data (e.g. MapReduce
	and GraphLab)
	6. working with message passing frameworks
	7. working with GPU/ coprocessor-machine
Submodules	none
Literature	
	• Anand Rajaraman, Jure Leskovec, and Jeffrey Ullman: Mining of massive datasets
	• Yucheng Low, Joseph Gonzalez, Aapo Kyrola, Danny Bick-
	son, Carlos Guestrinand Joseph M. Hellerstein: Distributed
	GraphLab: A Framework for Machine Learning and Data
	Mining in the Cloud PVLDB. 2012
Requirements	none
Exam	colloquium and written summary
Recommended Term	MSc 2
Turn	every summer term
Duration	1 Semester
Use	• MSc AInf Variant: Data Analytics (fundamental track) – Compulsory Modules
Language	English
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Module: Data and Privacy Protection

Responsible	Prof. Dr. Lars Schmidt-Thieme
Type	2 HPW lecture
Credit Points	3 CPs
Workload	presence: 21 hours; self-study: 54 hours
Learning goals/ Compe-	Students have an overview of specific requirements of data and pri-
tencies	vacy protection in different application areas such as e-commerce
00120100	and medicine. Students know basic laws about data and privacy
	protection in Germany, the EU and the US and can apply them to
	specific situations. Students have a broad overview of technological
	tools to protect data and privacy.
Content	The lecture provides an overview of methods for data and privacy
	protection, esp.
	1. Requirements of data and privacy protection
	a) General requirements
	b) Requirements in e-commerce
	c) Requirements in medicine
	2. Laws about data and privacy protection
	a) German Laws
	b) EU Laws
	c) US Laws
	3. Data and privacy protection policies & technologies
	a) IT security
	b) Data encryption c) Authorization and Rights Management
	C) Authorization and rughts Management
Submodules	none
Literature	
	• David G. Hill: Data Protection: Governance, Risk Manage-
	ment, and Compliance, Crc Pr Inc, 2009.
	• Helen Nissenbaum: Privacy in Context: Technology, Policy,
	and the Integrity of Social Life, Stanford Univ Pr, 2009.
Requirements	none
Exam	written exam (60 - 90 min) or oral exam (20 min)
Recommended Term	MSc 2
Turn	every summer term
Duration	1 Semester
Use	
	MSc AInf Variant: Data Analytics (fundamental track) –
	Compulsory Modules
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Language	English
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Module: Seminar Data Science II

Responsible	Prof. Dr. Niels Landwehr, Prof. Dr. Dr. Lars Schmidt-Thieme
Responsible Instructors	Prof. Dr. Dr. Lars Schmidt-Thieme und Angestellte der Arbeits-
	gruppe
Type	2 HPW Seminar
Credit Points	4 CPs
Workload	presence: 21 hours; self-study: 79 hours
Learning goals/ Competencies Content	Ziel des Seminars ist die selbständige Erschließung und Ausarbeitung eines vorgegebenen Inhaltsbereichs. Durch die Teilnahme an der Veranstaltung und die wissenschaftliche Diskussion über die Vorträge wird das Verständnis der bereits erworbenen Kenntnisse vertieft. Erwerb von Methodenkompetenzen, insb. besitzen Studierende eine Transferkompetenz, die es ihnen erlaubt, ihren Wissensstand selbstätig technischen und gesellschaftlichen Entwicklungen anzupassen. Generell werden einfache bis fortgeschrittene Themen behandelt, welche gewisse Grundlagen in Data Science voraussetzen. Ausgewählte Themen aus den Bereichen der Data Science. Beispiele für Seminarthemen: 1. Machine Learning 2. Data Mining 3. Data Visualization
Submodules	keine
Literature	je nach Thema
Requirements	Die Inhalte der Module "Maschinelles Lernen" und "Seminar Data Science I" werden vorausgesetzt.
Exam	Vortrag und Ausarbeitung. Die Sprache des Moduls und der Prüfung ist Deutsch oder Englisch.
Recommended Term	BSc 4-6
Turnus	jedes Semester
Duration	1 Semester
Use	• MSc AInf Variant: Data Analytics (fundamental track) – Compulsory Modules
Language	Englisch

Module: Seminar Data Analytics I

Responsible	Prof. Dr. Niels Landwehr, Prof. Dr. Dr. Lars Schmidt-Thieme
Type	2 HPW seminar
Credit Points	4 CPs
Workload	presence: 21 hours; self-study: 79 hours
Learning goals/ Competencies	The students deepen their analytic and methodological skills for understanding current research areas. Writing a report and giving a presentation as well as discussing scientific issues with their peers help the students to put the knowledge acquired during their studies into context and gives them an opportunity to add new knowledge to their corpus. Furthermore, skills are developed which will allow the students to adapt their knowledge to changing technical and societal conditions in the future.
Content	Selected topics in the area of data analytics. Covers advanced topics that require an advanced background in data science.
Submodules	SM 1: Seminar Data Analytics I Type: Seminar (4 CPs) Lecturer: Prof. Dr. Dr. Lars Schmidt-Thieme and members of the study group
Literature	depending on the topic
Requirements	none
Exam	Presentation and written summary
Recommended Term	MSc 1-3
Turn	every semester
Duration	1 Semester
Use	• MSc AInf Variant: Data Analytics (fundamental track) – Compulsory Modules
Language	English

Module: Seminar Data Analytics II

Responsible	Prof. Dr. Niels Landwehr, Prof. Dr. Dr. Lars Schmidt-Thieme
Type	2 HPW seminar
Credit Points	4 CPs
Workload	presence: 21 hours; self-study: 79 hours
Learning goals/ Competencies	The students deepen their analytic and methodological skills for understanding current research areas. Writing a report and giving a presentation as well as discussing scientific issues with their peers help the students to put the knowledge acquired during their studies into context and gives them an opportunity to add new knowledge to their corpus. Furthermore, skills are developed which will allow the students to adapt their knowledge to changing technical and societal conditions in the future.
Content	Selected topics in the area of data analytics. Covers very advanced topics that require a very advanced background in data science.
Submodules	SM 1: Seminar Data Analytics II Type: Seminar (4 CPs) Lecturer: Prof. Dr. Lars Schmidt-Thieme and members of the study group
Literature	depending on the topic
Requirements	Seminar Data Analytics I
Exam	Presentation and written summary
Recommended Term	MSc 1-3
Turn	every semester
Duration	1 Semester
Use	• MSc AInf Variant: Data Analytics (fundamental track) – Compulsory Modules
Language	English

Module: Student Research Project

Responsible	Prof. Dr. Niels Landwehr, Prof. Dr. Dr. Lars Schmidt-Thieme
Type	4 HPW project
Credit Points	15 CPs
Workload	presence: 42 hours; self-study: 333 hours
Learning goals/ Compe-	Students independently learn how to plan and manage their
tencies	projects and thus develop methodological and social skills.
	Through teamwork they develop social skills such as conflict solv-
	ing strategies, communicative skills and team management. In this
	project students improve their experience in research methodology,
	they independently formulate research topics, create the research
	design and reflect on them critically and organize the implemen-
	tation of research processes. They should be in a position to in-
	dependently work on a new topic and use known methods on new
	problems. They are able to document their project and present
	their solution.
Content	In teams students work out a contribution for a extensive problem.
	The following contents are treated, independently from the specific
	task:
	1. Iterative formulation, verification and revision of research
	topics
	2. Iterative design and critical reflection of research design
	3. Basic and advanced research methods
	4. Project management methods
	5. Organization, management, leading group sessions
	6. Project documentation
	Other contents depends on the project task.
Submodules	SM 1: Student Research Project, Part I Type: Project (6 CPs)
	Lecturer: Prof. Dr. Dr. Lars Schmidt-Thieme SM 2: Student
	Research Project, Part II Type: Project (9 Cps) Lecturer: Prof.
	Dr. Dr. Lars Schmidt-Thieme and members of the study group
Literature	depending on the topic
Requirements	none
Exam	Project
Recommended Term	MSc 2-3
Turn	every semester
Duration	2 Semester
Use	
	• MSc AInf Variant: Data Analytics (fundamental track) –
	Compulsory Modules
Language	English
Danguage	Tuguon

Elective Modules

Methodological Specialization

Module: Advanced Case-Based Reasoning

Responsible	Dr. Pascal Reuss
Type	2 HPW lecture, 2 HPW tutorial
Credit Points	6 CPs
Workload	presence: 42 hours; self-study: 108 hours
Learning goals/ Competencies	Students have an extended understanding of case-based reasoning. They master advanced and detailed procedures to develop, operate and maintain case-based reasoning and be able to use them for more complex scenarios. Students are able to assign complex scenarios and case studies to special task classes and to state-of-the-art and state-of-the-practice.
Content	Development, operation and maintainance of case-based reasoning and its application. Some characteristics of case-based reasoning like case-based classification, diagnosis and decision making, configuration and design and case-based planning are presented for special task categories. The application potential is shown in case studies and in state-of-the-art/practice-systems.
Submodules	SM 1: Advanced Case-Based Reasoning, Lecture Type: 2 HPW Lecture (3 CPs) Lecturer: Dr. Pascal Reuss SM 2: Advanced Case-Based Reasoning, Tutorium Type: 1 HPW Tutorium (2 CPs) Lecturer: Prof. Dr. Klaus-Dieter Althoff and members of the study group
Literature	
	 M.M. Richter, R.O. Weber: Case-Based Reasoning, Springer, Berlin 2013 R. Bergmann: Experience Management- Foundations, Development Methodology, and Internet-Based Applications. Springer, Berlin 2002. R. Bergmann, KD. Althoff, S. Breen, M. Göker, M. Manago, R. Traphöner, S. Wess: Developing Industrial Case-Based Reasoning Applications - The INRECA Methodology. Springer, Berlin 2003. M. Lenz, B. Bartsch-Spörl, HD. Burkhard, S. Wess (Hrsg.): Case-Based Reasoning Technology: From Foundations to Applications. Springer, Berlin 1998.
Requirements	none
Exam	written exam (90 minutes)
Recommended Term	MSc 1-3

Turn	irregular turn, for additional information on next lecture turn
	please contact Dr. Pascal Reuss
Duration	1 Semester
Use	MSc AInf Variant: Data Analytics (fundamental track) – Elective Modules – Methodological Specialization
Language	English

Module: Advanced Computer Vision

Responsible	Prof. Dr. Niels Landwehr
Responsible Instructors	See submodules
Type	2 HPW lecture, 2 HPW tutorial
Credit Points	6 CPs
Workload	presence: 42 hours; self-study: 108 hours
Learning goals/ Compe-	After completion of the module, students are familiar with different
tencies	problem settings and tasks studied in modern computer vision. They understand how machine learning techniques and specifically deep neural networks can be used to solve these tasks. They are able to build upon established techniques and adapt them to novel problem settings, enabling them to solve practical computer vision problems.
Content	The lecture starts with an overview of deep neural networks. It then discusses design principles of deep neural network architectures for computer vision problems. Specific computer vision problems such as image classification, segmentation, object detection and localization, or metric learning are discussed. The lecture also presents examples for practical applications of computer vision techniques in different domains.
Submodules	SM 1: Advanced Computer Vision, lecture Type: 2 HPW lecture (3 CPs) Lecturer: Prof. Dr. Niels Landwehr SM 2: Advanced Computer Vision, tutorial Type: 2 HPW tutorial (3 CPs) Lecturer: Prof. Dr. Niels Landwehr and members of the study group
Literature	
	 Goodfellow, Ian, Bengio, Yoshua, Courville, Aaron. (2016). Deep learning. Cambridge: MIT press. Szeliski, Richard. Computer vision: algorithms and applications. Springer Science & Business Media, 2010.
Requirements	Completion of the module "Deep Learning" or a simultaneous en- rollment in the module "Deep Learning" is recommended.
Exam	written exam (90 minutes) or oral exam (30 min), can also take place online.
Recommended Term	MSc 1-3
Turn	winter semester
Duration	1 semester
Use	• MSc AInf Variant: Data Analytics (fundamental track) – Elective Modules – Methodological Specialization
Language	English

Module: Bayesian Networks

Responsible	Prof. Dr. Lars Schmidt-Thieme
Type	2 HPW lecture, 2 HPW tutorial
Credit Points	6 CPs
Workload	presence: 42 hours; self-study: 108 hours
Learning goals/ Competencies Content	Students have detailed knowledge about Bayesian Networks. They are able to model problems using Bayesian Networks. They understand exact and approximative inference methods and are able to choose suitable methods depending on the problem. Students know learning methods for parameter and structure and can estimate the results of learning processes. They get used to work with new books in Bayesian Networks. The lecture introduces bayesian networks. Based on modelling influences and conditional probabilities, algorithms for exact and approximative inference, analysis of bayesian networks, learning parameters and learning structure are treated. Algorithms for infer-
	ence and learning bayesian networks go back to graph-algorithms as well as methods like topological sorting and connectivity-property and specific methods like counting cliques and so on. Every necessary algorithm will be introduced in the lecture.
Submodules	SM 1: Bayesian Networks, Lecture Type: 2 HPW Lecture (3 CPs) Lecturer: Prof. Dr. Dr. Lars Schmidt-Thieme SM 2: Bayesian Networks, Tutorium Type: 2 HPW Tutorium (3 CPs) Lecturer: Prof. Dr. Dr. Lars Schmidt-Thieme and members of the study group
Literature	
	 Marco Scutari: Bayesian Networks: With Examples in R, Chapman and Hall/CRC, 2014. D. Koller, N. Friedmann: Probabilistic Graphical Models: Principles and Techniques, The MIT Press, 2009. Finn V. Jensen: Bayesian networks and decision graphs. Springer, 2001. Richard E. Neapolitan: Learning Bayesian Networks. Prentice Hall, 2003. Enrique Castillo, Jose Manuel Gutierrez, Ali S. Hadi: Expert Systems and Probabilistic Network Models. Springer, 1997. Christian Borgelt, Rudolf Kruse: Graphical Models. Wiley, 2002.
Requirements	none
Exam	written exam (60 - 90 min) or oral exam (20 min)
Recommended Term	MSc 1-3
Turn	not regularly, normally every 4th semester
Duration	1 Semester
Exam Recommended Term	none written exam (60 - 90 min) or oral exam (20 min)

Use	• MSc AInf Variant: Data Analytics (fundamental track) – Elective Modules – Methodological Specialization
Language	English

Module: Business Analytics

Responsible	Prof. Dr. Lars Schmidt-Thieme
Lecturer	none
Type	2 HPW lecture, 2 HPW tutorial
Credit Points	6 CPs
Workload	presence: 42 hours; self-study: 108 hours
Learning goals/ Compe-	
tencies	 Understanding the classical forecasting methodologies and their application to business domains Exploring the state-of-the-art in terms of Recommender Systems and the Internet economy Empowering the analytical ability to abstract the necessary data-driven methodologies for complex business problems
Content	Business Analytics aims at introducing students to the fundamental data science know-how, which provides a start-level proficiency for tackling data-driven business problems. Initially the course explains prediction models for Regression and Classification tasks, as well as typical Clustering approaches. Frequent Pattern Mining that discovers association rules from transactional data will be covered as well. Dimensionality Reduction techniques are taught with regards to both visualisation and feature extraction aspects. In addition, personalized strategies in the realm of Recommender Systems will be exploited. On the other hand, the course covers Time-Series Forecasting methods, as well as Process Mining from industrial data logs. Last, but not least, the course aims at providing an introduction on current strategies needed to scale data analytics methods to handle big data.
Submodules	SM 1: Business Analytics, Lecture Type: 2 HPW Lecture (3 CPs) Lecturer: Prof. Dr. Dr. Lars Schmidt-Thieme SM 2: Business Analytics, Tutorium Type: 2 HPW Tutorium (3 CPs) Lecturer: Prof. Dr. Dr. Lars Schmidt-Thieme and members of the study group
Literature	
	 Hyndman et al., Forecasting: Principles and Practice, 2012 Aggarwal et al., Frequent Pattern Mining, 2014 Aggarwal, Recommender Systems, 2016 Tie-Yan Liu, Learning to Rank for Information Retrieval, 2011
Requirements	none
Exam	written exam (120 min) or an oral exam (30 min)
Recommended Term	MSc 1
Turn	every winter term
Duration	1 Semester
	1 Somester

Use	MSc AInf Variant: Data Analytics (fundamental track) – Elective Modules – Methodological Specialization
Language	English

Module: Computer Vision

Responsible	Prof. Dr. Lars Schmidt-Thieme
Type	2 HPW lecture, 2 HPW tutorial
Credit Points	6 CPs
Workload	presence: 42 hours; self-study: 108 hours
Learning goals/ Competencies	After the completion of this module, the students should be able to solve relevant tasks and research questions from the Computer Vision. They should have developed a deeper understanding in the field of Computer Vision. They should be able to understand, implement and apply different Computer Vision techniques. The students should be capable of adapting those techniques to specific applications. In addition, they should be in a position to understand and elaborate further procedures based on the literature.
Content	The course will cover statistical data-driven approaches for automatic processing, analyzing and understanding of images. The lecture will cover topics from the pre-processing of images, like image filtering and feature detection to object recognition and object tracking as well as image classification.
Submodules	SM 1: Computer Vision, Lecture Type: 2 HPW Lecture (3 CPs) Lecturer: Prof. Dr. Dr. Lars Schmidt-Thieme SM 2: Computer Vision, Tutorium Type: 2 HPW Tutorium (3 CPs) Lecturer: Prof. Dr. Dr. Lars Schmidt-Thieme and members of the study group
Literature	
	 Richard Szeliski: Computer Vision: Algorithms and Applications. Microsoft Research, 2010. Milan Sonka, Vaclav Hlavac, Roger Boyle: Image Processing, Analysis, and Machine Vision. Thomson, 2008. John C. Russ, J. Christian Russ: Introduction to Image Processing and Analysis. CRC Press, 2008. R. C. Gonzalez, R. E Woods: Digital Image Processing. Pearson, 2008. G. Aubert, P. Kornprobst: Mathematical Problems in Image Processing. Partial Differential Equations and the Calculus of Variations. Springer, 2006. J. R. Parker: Algorithms for Image Processing and Computer Vision. Wiley, 1997.
Requirements	none
Exam	written exam (120 min) or an oral exam (30 min)
Recommended Term	MSc 1-3
Turn	not regularly, normally every 4th semester
Duration	1 Semester
Use	• MSc AInf Variant: Data Analytics (fundamental track) – Elective Modules – Methodological Specialization
Language	English

Module: Constraint Logic Programming

Responsible	Prof. Dr. Klaus Schmid
Responsible Instructors	Prof. Dr. Klaus Schmid and members of the study group
Type	2 HPW lecture and tutorial
Credit Points	3 CPs
Workload	presence: 21 hours; self-study: 54 hours
Learning goals/ Competencies	Students learn the basics of alternative programming paradigms using the examples of constraint programming and logic programming. The students will learn the basic theoretical concepts and application examples, which help them understand the advantages and disadvantages of these programming techniques and their limitations. As a result, students can use these approaches in a goal-oriented and thoughtful way to solve concrete problems and classify these approaches' suitability.
Content	The basic concepts of constraint programming and logical programming will be presented. Different classes of constraint programming languages and the corresponding constraint systems will be discussed. Concrete languages and solvers are described and used in the exercises. The special cases of certain domains like Boolean algebra or finite domains will be explained. The courses each contain an exercise part to deepen the respective ideas.
Literature	 Thom Frühwirth, Slim Abdennadher. Essentials of Constraint programming, Springer 2003 Krzysztof Apt. Principles of Constraint Programming, Cambridge University Press, 2003
Requirements	none
Exam	written exam (60 - 90 min) or oral exam (20 min)
Recommended Term	MSc 1-3
Turn	non-regular
Duration	1 Semester
Use	 MSc AInf Variant: Data Analytics (fundamental track) – Elective Modules – Methodological Specialization – Computer Vision MSc AInf Variant: Data Analytics (fundamental track) – Elective Modules – Application – Computer Science/Software Engineering
Language	English

Module: Deep Learning

Responsible	Prof. Dr. Niels Landwehr, Prof. Dr. Dr. Lars Schmidt-Thieme
Type	2 HPW lecture, 2 HPW tutorial
Credit Points	6 CPs
Workload	presence: 42 hours; self-study: 108 hours
Learning goals/ Compe-	Deep learning has recently been associated with revolutionary Ar-
tencies	tificial Intelligence achievements, ranging from "close-to-human"
	speech and image recognition performances, up to "super-human"
	game playing results. Throughout this course, students will have
	the opportunity to understand the building blocks of neural net-
	works
Content	The curriculum starts by introducing supervised learning concepts
	and incrementally dives into the peculiarities of learning the pa-
	rameters of neural networks through back-propagation. Specific ar-
	chitectures, such as the Convolutional Neural Networks will be cov-
	ered, as well as different types of network regularization strategies.
	Furthermore implementation techniques involving GPU-based op-
	timization will be explained. The students are expected to master
	the necessary knowledge that will empower them to apply Deep
	Learning in real-life problems.
Submodules	SM 1: Deep Learning, lecture Type: 2 HPW lecture (3 CPs) Lec-
	turer: Prof. Dr. Dr. Lars Schmidt-Thieme SM 2: Deep Learning,
	tutorial Type: 2 HPW tutorial (3 CPs) Lecturer: Prof. Dr. Dr.
	Lars Schmidt-Thieme and members of the study group
Literature	will be announced in the lecture
Requirements	none
Exam	written exam (120 min) or an oral exam (30 min)
Recommended Term	MSc 2
Turn	every 4th Semester but not regularly.
Duration	1 Semester
Use	
	MSc AInf Variant: Data Analytics (fundamental track) –
	Elective Modules – Methodological Specialization
Language	English

Module: Deep Learning 2

Responsible	Prof. Dr. Lars Schmidt-Thieme
Responsible Instructors	siehe Teilmodulbeschreibungen
Type	2 HPW lecture, 2 HPW tutorial
Credit Points	6 CPs
Workload	presence: 42 hours; self-study: 108 hours
Learning goals/ Compe-	Deep learning has recently been associated with revolutionary Ar-
tencies	tificial Intelligence achievements, ranging from "close-to-human"
	speech and image recognition performances, up to "super-human"
	game playing results. Throughout this course, students will have
	the opportunity to understand the building blocks of neural networks
Content	Based on the major concepts studied in the introductory lecture
Content	Deep Learning, this lecture will study advanced topics in deep
	learning such as
	• equivariant and invariant network architectures
	• differentiable architecture search
	• meta learning
	• few shot learning
	• multi task learning
	• hyperparameter optimization for deep neural networks
Submodules	SM 1: Deep Learning 2, Lecture Type: 2 HPW Lecture (3 CPs) Lecturer: Prof. Dr. Dr. Lars Schmidt-Thieme SM 2: Deep Learn- ing 2, Tutorium Type: 2 HPW Lecture (3 CPs) Lecturer: Prof. Dr. Dr. Lars Schmidt-Thieme und Angestellte der Arbeitsgruppe
Literature	
	 Aston Zhang, Zachary C. Lipton, Mu Li, and Alexander J. Smola. Dive into Deep Learning. 2020. https://d2l.ai current papers
Requirements	keine
Exam	written exam (120 min) or an oral exam (30 min) Die Sprache des Moduls und der Prüfung ist Englisch.
Recommended Term	MSc 1-3
Turn	jedes Sommersemester
Duration	1 Semester
Use	
Cisc	• MSc AInf Variant: Data Analytics (fundamental track) – Elective Modules – Methodological Specialization
Language	Englisch

Module: Deep Learning Masterclass

Responsible	Prof. Dr. Lars Schmidt-Thieme
Responsible Instructors	Prof. Dr. Artus Krohn-Grimberghe
Type	4 HPW lecture, 4 HPW tutorial
Credit Points	6 CPs
Workload	presence: 84 hours; self-study: 66 hours
Learning goals/ Compe-	The course allows students to gain practical knowledge and capa-
tencies	bilities in the area of Deep Learning. Students will be able to reproduce probabilistic models from state-of-the-art techniques from recent papers.
Content	This 10-day full-time instructor-led deep-dive course for coders consists of three parts. The first part, an overview over deep learning and deep neural networks, which problems they are applicable to, how they work and how they are implemented on a very high level (using pytorch building blocks to be precise) on day 1. The second part teaches how to code deep learning using deep neural networks efficiently for various problem settings such as image classification, multi-class classification, tabular data, audio, image segmentation, superresolution, neural style transfer, GAN and NLP on days 2-6. The third part re-creates large parts of fast.ai and pytorch as an optional module for those who want to dive deep into the inner workings of deep learning during days 7-10. The days of this course are structured such that the mornings consist of recorded lectures presenting the jupyter notebooks with the course contents and the afternoons consist of paper reading and presentation groups (reading several of the original seminal and brand new publication that drive the field), code presentation groups and guided coding and q&a sessions. Participants are encouraged to apply the learned content on their own datasets or rehearse or prepare materials during the evenings.
Submodules	none
Literature	will be announced in the lecture
Requirements	none
Exam	Project and written report
Recommended Term	MSc 2
Turn	Every Semester
Duration	1 Semester
Use	• MSc AInf Variant: Data Analytics (fundamental track) – Elective Modules – Methodological Specialization
Language	English

Module: Graph Analytics

Responsible	Prof. Dr. Jürgen Sander
Responsible Instructors	Dr. Jan-Hendrik de Wiljes
Type	2 HPW lecture, 2 HPW tutorial
Credit Points	6 CPs
Workload	presence: 42 hours; self-study: 108 hours
Learning goals/ Competencies	The students will be able to model and analyse (static) network structures with graphs and graph theoretic methods. Further, they get an understanding of network dynamics and corresponding models like random graphs and percolation models. Finally, the students use graphs to model (general) data and solve typical problems via the application of state-of-the-art methods.
Content	The following aspects will be presented and discussed: 1. Introduction, network classification and typical problems 2. Graph partitioning and community detection 3. Power and centrality analysis 4. Link analysis (and link prediction) 5. Network dynamics (and models) 6. Modelling "non-graphic" data (similarity graphs,)
Submodules	none
Literature	• will be announced in the lecture
Requirements	none
Exam	written exam (120 min) or an oral exam (30 min)
Recommended Term	MSc 2-4
Turn	No longer offered.
Duration	1 Semester
Use	• MSc AInf Variant: Data Analytics (fundamental track) – Elective Modules – Methodological Specialization
Language	Englisch

Module: Lab Course Deep Learning

Type Credit Points 6 CPs Workload presence: 42 hours; self-study: 108 hours Learning goals/ Competencies The lab allows students to gain practical knowledge and capabilities in the area of Deep Learning. Students will be able to reproduce probabilistic models from state-of-the-art techniques from recent papers. Content The lab allows students to gain practical knowledge and capabilities in the area of Deep Learning. This implementation-oriented course offers hands-on experience with current algorithms and approaches in Deep Learning and their application to real-world learning and decision-making tasks. This course will provide expabilities for students to reproduce experiments seen in papers and also how to model their discoveries. This course also aims to guide students in how to use Deep Learning tools and also to adopt healthy implementations practices. The methods being taught will change according to recent publications allowing students to participate in research in current topics. Methods will include: 1. Basic usage of Deep Learning Tools 2. Implementation of basic types of networks: CNN, RNN, and FCN. 3. Data Pre-Processing 4. Image Classification 5. Image Segmentation 6. Time Series Analysis 7. Recommended Systems Submodules none Literature will be announced in the lab Deep Learning Lecture Exam colloquium and written summary Recommended Term MSc 2 • MSc Alm Variant: Data Analytics (fundamental track) – Elective Modules – Methodological Specialization	Responsible	Prof. Dr. Lars Schmidt-Thieme
Description Description	Type	4 HPW lab course
Learning goals/ Competencies The lab allows students to gain practical knowledge and capabilities in the area of Deep Learning. Students will be able to reproduce probabilistic models from state-of-the-art techniques from recent papers. Content The lab allows students to gain practical knowledge and capabilities in the area of Deep Learning. This implementation-oriented course offers hands-on experience with current algorithms and approaches in Deep Learning and their application to real-world learning and decision-making tasks. This course will provide capabilities for students to reproduce experiments seen in papers and also how to model their discoveries. This course also aims to guide students in how to use Deep Learning tools and also to adopt healthy implementations practices. The methods being taught will change according to recent publications allowing students to participate in research in current topics. Methods will include: 1. Basic usage of Deep Learning Tools 2. Implementation of basic types of networks: CNN, RNN, and FCN. 3. Data Pre-Processing 4. Image Classification 5. Image Segmentation 6. Time Series Analysis 7. Reccomender Systems Submodules none Literature will be announced in the lab Deep Learning Lecture Exam colloquium and written summary Recommended Term MSc 2 Turn every summer term Duration 1 Semester • MSc Alnf Variant: Data Analytics (fundamental track) — Elective Modules — Methodological Specialization	Credit Points	6 CPs
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5. Image Segmentation 6. Time Series Analysis 7. Reccomender Systems Submodules In the lab Requirements Deep Learning Lecture Exam Colloquium and written summary Recommended Term MSc 2 Turn Duration 1 Semester • MSc AInf Variant: Data Analytics (fundamental track) – Elective Modules – Methodological Specialization		
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Recommended Term MSc 2 Turn every summer term Duration 1 Semester Use MSc AInf Variant: Data Analytics (fundamental track) – Elective Modules – Methodological Specialization	Requirements	
Turn every summer term Duration 1 Semester Use • MSc AInf Variant: Data Analytics (fundamental track) – Elective Modules – Methodological Specialization	Exam	colloquium and written summary
Duration 1 Semester Use • MSc AInf Variant: Data Analytics (fundamental track) – Elective Modules – Methodological Specialization		MSc 2
Use • MSc AInf Variant: Data Analytics (fundamental track) – Elective Modules – Methodological Specialization		· ·
• MSc AInf Variant: Data Analytics (fundamental track) – Elective Modules – Methodological Specialization		1 Semester
Elective Modules – Methodological Specialization	Use	MSc AInf Variant: Data Analytics (fundamental track) –
Language English		_ ` ` ` ` ` ` ` ` ` ` ` ` ` ` ` ` ` ` `
	Language	English

Module: Machine Learning for IT Security

Responsible	Prof. Dr. Niels Landwehr
Responsible Instructors	See submodules
Type	2 HPW lecture, 2 HPW tutorial
Credit Points	6 CPs
Workload	presence: 42 hours; self-study: 108 hours
Learning goals/ Compe-	After completion of the module, students have a detailed under-
tencies	standing of how machine learning can be used in IT security to
	detect and counter threats and attacks. They are able to map
	IT security problems to appropriate problem settings in machine
	learning and choose effective data representations for a given prob-
	lem. They also understand the limitations of using data-driven
	methods such as machine learning in a security context. They are
	finally able to read and follow the current literature on machine
	learning in IT security to further enhance their knowledge about
Content	the topic. The lecture studies different threats and tasks in IT security (such
Content	as filtering malicious email messages, detecting malicious exe-
	cutable files, discovering security vulnerabilities in source code,
	or detecting fraudulent activity). We discuss how such tasks can
	be cast as machine learning problems, the process of data collec-
	tion and data representation, and appropriate machine learning
	techniques for solving these tasks.
Submodules	SM 1: Machine Learning for IT Security, lecture Type: 2 HPW
	lecture (3 CPs) Lecturer: Prof. Dr. Niels Landwehr SM 2: Ma-
	chine Learning for IT Security, tutorial Type: 2 HPW tutorial (3
	CPs) Lecturer: Prof. Dr. Niels Landwehr and members of the
	study group
Literature	
	• Salomon, David. Elements of computer security. Springer
	Science & Business Media, 2010.
	• Tom Mitchell: Machine Learning. McGraw-Hill, 1997.
	• Thomas, Tony, Athira P. Vijayaraghavan and Sabu Em-
	manuel. Machine Learning Approaches in Cyber Security Analytics. Springer, 2020.
	Analytics. Springer, 2020.
	Further literature will be announced in the lecture
Requirements	Some prior knowledge in the area of machine learning is recom-
	mended
Exam	written exam (90 minutes) or oral exam (30 min), can also take
	place online.
Recommended Term	BSc 5, MSc 1-3
Turn	summer semester
Duration	1 semester

Us	se	• MSc AInf Variant: Data Analytics (fundamental track) – Elective Modules – Methodological Specialization
La	anguage	English

Module: Survey Sampling

Responsible	PD Dr. Jürgen Groß
Responsible Instructors	PD Dr. Jürgen Groß
Type	2 HPW lecture, 2 HPW tutorial
Credit Points	6 CPs
Workload	presence: 42 hours; self-study: 108 hours
Learning goals/ Compe-	Students should be able to discern different elementary sampling
tencies	designs, understand and handle the statistical implications of such designs, and be able to apply an appropriate sampling design to retrieve information about a target population. In addition, students should acquire knowledge of a statistical programming language to implement specific sampling designs tailored to accommodate given data.
Content	The course gives an introduction to the statistical theory and methods for sample surveys based on probability sampling. Some elementary sampling designs are presented and their statistical and mathematical implications are discussed. The treated sampling designs include: • Bernoulli Sampling, Simple Random Sampling, Systematic Sampling, Poisson Sampling, Probability Proportional-to-Size Sampling, Stratified Sampling, Cluster Sampling.
Submodules	none
Literature	 Särndal, CE., Swensson, B. & Wretmann, J. (1992). Model Assisted Survey Sampling. Springer. Thompson, S.K. (2012). Sampling. Third edition. Wiley.
Requirements	none
Exam	written exam (120 min) or an oral exam (30 min)
Recommended Term	MSc 1-3
Turn	every winter term
Duration	1 Semester
Use	• MSc AInf Variant: Data Analytics (fundamental track) – Elective Modules – Methodological Specialization
Language	Englisch

Module: Time Series Analysis

Responsible	Prof. Dr. Sebastian Mentemeier
Type	2 HPW lecture, 2 HPW tutorial
Credit Points	6 CPs
Workload	presence: 42 hours; self-study: 108 hours
Learning goals/ Competencies	After completion of this module, students are able to describe and analyse time series data with methods from probability theory and statistics, to make predictions about future development of the observed time series and to make decisions based on observations of time series data. The students have a deeper understanding of mathematical models for time series and their fields of applications. They have good command of a statistical programming language, e.g. R.
Content	This lecture gives an overview of time series analysis from the viewpoint of mathematical statistics. It focusses on:
	 Decomposition: Identification of trends and seasonal components Models for discrete time series: Autoregressive Models, Moving Average Models and ARMA models, parameter estimation for these models Models for heteroskedasticity: ARCH and GARCH models and parameter estimation Aspects of extreme value theory for time series Models for continuous time series: Brownian motion and related stochastic processes - if time permits.
Submodules	SM 1: Time Series Analysis, Lecture Type: 2 HPW Lecture (3 CPs) Lecturer: Prof. Dr. Sebastian Mentemeier SM 2: Time Series Analysis, Tutorium Type: 2 HPW Tutorium (3 CPs) Lecturer: Prof. Dr. Sebastian Mentemeier and members of the study group
Literature	
	 P. Cowpertwait, A. Metcalfe: Introductory Time Series with R, Springer 2009 P. Brockwell, R. Davis: Introduction to Time Series and Forecasting, Springer 1996
Requirements	none
Exam	written exam (120 min) or an oral exam (30 min)
Recommended Term	MSc 1-3
Turn	Every summer semester.
Duration	1 Semester
Use	• MSc AInf Variant: Data Analytics (fundamental track) – Elective Modules – Methodological Specialization
Language	English

Application

Computer Science/Software Engineering

Module: Software Engineering

Responsible	Prof. Dr. Klaus Schmid
Responsible Instructors	Prof. Dr. Klaus Schmid and members of the study group
Type	2 HPW lecture, 2 HPW tutorial
Credit Points	6 CPs
Workload	presence: 42 hours; self-study: 108 hours
Learning goals/ Compe-	The aim of this module is to provide a basic understanding of
tencies	the problems, challenges and solution approaches of software engineering. The focus is on teaching essential techniques, as well as the methodological approaches of systematic software development. Students will acquire competencies for problem analysis, as well as competencies in the area of design and implementation of IT systems. In particular, students understand the fundamentals of software engineering workflows and software engineering modelling techniques. The fundamentals of software development are taught on a large
Content	scale. This includes in particular:
	 Process, product and quality reference models Process models and life cycle models such as the waterfall model, spiral model Requirements engineering (among other things, use cases, business process modeling) Software architecture (among other things, architecture styles, design patterns) Implementation techniques Testing techniques (black-box, white-box) Verification techniques (formal verification, inspection techniques) In the tutorial, the contents taught in the lecture are deepened by manne of everyings. In the course of the everying techniques
	means of exercises. In the course of the exercise, tasks are worked on together as well as homework is distributed and corrected. The focus is on teaching the students the competence of independently applying the discussed techniques.
Literature	
	 Ian Sommerville, Software Engineering, 10th Edition. Pearson, 2015. Current scientific publications. Announced in class. Additional Material, made available in class.
Requirements	general knowledge of computer science
Exam	written exam (120 min) or an oral exam (30 min)
Recommended Term	BSc 4 or MSc 1
Turn	every winter term

Duration	1 Semester
Use	• MSc AInf Variant: Data Analytics (fundamental track) - Elective Modules - Application - Computer Science/Software Engineering
Language	English

Module: Model-based Software Engineering

Responsible	Dr. Holger Eichelberger
Responsible Instructors	Dr. Holger Eichelberger; Prof. Dr. Klaus Schmid and members of
	the study group
Type	2 HPW lecture, 2 HPW tutorial
Credit Points	6 CPs
Workload	presence: 42 hours; self-study: 108 hours
Learning goals/ Compe-	Students are able to name the essential differences between various
tencies	types of models in software development, to describe transformations between models and to reflect on these in the context of given use cases. They know the current state of science in this area and are able to relate different approaches to each other, or to distinguish them from each other. They know the current state of science in this area and are able to classify current work. They are able to continuously develop their knowledge.
Content	The creation, processing, and analysis of models will be presented. Current approaches in these areas will be discussed. Main topics of the lecture are:
	1. Formal basics of models
	2. Meta-modeling
	3. Model semantics
	4. Textual and graphical models
	5. Model-to-model transformations6. Model-to-text transformations
	In the exercise, the contents taught in the lecture Model-Based Development are deepened by means of exercises. Tasks will be worked on together in the exercise and questions concerning the analysis of the methods will be discussed. The focus is on teaching the competence for independent analysis and further development by the students.
Literature	
	 Current scientific publications. Announced in class. Additional Material, made available in class.
Requirements	Knowledge of Software Engineering as thought, for example, in the course Software Engineering.
Exam	written exam (120 min) or an oral exam (30 min)
Recommended Term	MSc 1-3
Turn	every winter term
Duration	1 Semester
Use	• MSc AInf Variant: Data Analytics (fundamental track) – Elective Modules – Application – Computer Science/Software Engineering
Language	English

Module: Advanced Web Development

the area of modern web technologies and architectures. They re ognize and understand the practical constraints relevant to the in plementation of modern web applications. They are able to reflect on possibilities and application areas of web technologies as well associated development environments and to assess them in context. They acquire the technical competences for the realissation of an exemplary web application, especially by working in a teat and the technical competences for the realissation of an exemplary web application, especially by working in a teat and the technical skills, technology and effectiveness assessment. Content In this course, students learn software development for current web technologies. For this purpose, the technical basics of website eation, current web technologies and frameworks, associated devopment environments and techniques are explained and practic in the form of an accompanying development project. The devopment work is done in a team and includes realization and te ing of an (in the final stage) platform-dependent web application and the ing of an (in the final stage) platform-dependent web application and the ing of an (in the final stage) platform-dependent web application and the ing of an (in the final stage) platform-dependent web application and the ing of an (in the final stage) platform-dependent web application and the ing of an (in the final stage) platform-dependent web application and the ing of an (in the final stage) platform-dependent web application and the ing of an (in the final stage) platform-dependent web application and the ing of an (in the final stage) platform-dependent web application and the ing of an (in the final stage) platform-dependent web application and the ing of an (in the final stage) platform-dependent web application and the ing of an (in the final stage) platform-dependent web application and the ing of an (in the final stage) platform-dependent web application and the ing of an (in the final stage) platform-dependent web appli	Responsible	Dr. Holger Eichelberger
Credit Points	Responsible Instructors	Dr. Holger Eichelberger and members of the study group
Workload presence: 42 hours; self-study: 108 hours	Type	2 HPW lecture, 2 HPW lab course
Learning goals/ Competencies Students acquire analytical and methodological competencies the area of modern web technologies and architectures. They represented by the area of modern web applications. They are able to refine the plementation of modern web applications. They are able to refine on possibilities and application areas of web technologies as well associated development environments and to assess them in context. They acquire the technical competences for the realisation of an exemplary web application, especially by working in a teat. Through teamwork and exchange with the lecturers, they acquire skills such as conflict resolution strategies, communication skills, technology and effectiveness assessment. Content In this course, students learn software development for current we technologies. For this purpose, the technical basics of website contents and techniques are explained and practic in the form of an accompanying development project. The development work is done in a team and includes realization and the ingoing of an (in the final stage) platform-dependent web applications to techniques and tools. The tasks typically relate to practicing the techniques and tools. The tasks typically relate to practicing the techniques and technologies explained. Literature Proven programming skills in at least one programming language • Knowledge of Software Engineering as thought, for examping the course Software Engineering. Exam Final presentation, prototype and written elaboration/documentation. Recommended Term MSc 1-3 Turn every winter term Duration 1 Semester • MSc AInf Variant: Data Analytics (fundamental trace Elective Modules – Application – Computer Software Engineering – Computer Software Engineering – Computer Software Elective Modules – Application – Computer Software Elective Modules – Application – Computer Software – Elective Modules – Application – Computer Software – Elective Modules – Application – Computer Software – Computer Software – Computer Software – Computer Softwar	Credit Points	6 CPs
the area of modern web technologies and architectures. They re ognize and understand the practical constraints relevant to the i plementation of modern web applications. They are able to refer on possibilities and application areas of web technologies as well associated development environments and to assess them in context. They acquire the technical competences for the realisation of an exemplary web application, especially by working in a teat Through teamwork and exchange with the lecturers, they acques social skills such as conflict resolution strategies, communicating skills, technology and effectiveness assessment. Content In this course, students learn software development for current web technologies. For this purpose, the technical basics of websites ation, current web technologies and frameworks, associated development environments and techniques are explained and practic in the form of an accompanying development project. The devolution of an (in the final stage) platform-dependent web application of an (in the final stage) platform-dependent web application of an (in the final stage) platform-dependent web application in the course sary activities and practice the underly techniques and tools. The tasks typically relate to practicing the techniques and tools. The tasks typically relate to practicing the techniques and technologies explained. Eiterature Proven programming skills in at least one programming leguage • Knowledge of Software Engineering as thought, for examping the course Software Engineering. Exam Final presentation, prototype and written elaboration/documentation. Recommended Term MSc 1-3 Turn every winter term Duration 1 Semester • MSc AInf Variant: Data Analytics (fundamental transpurse) and the plate of the plantation of	Workload	presence: 42 hours; self-study: 108 hours
technologies. For this purpose, the technical basics of website c ation, current web technologies and frameworks, associated devopment environments and techniques are explained and practic in the form of an accompanying development project. The devopment work is done in a team and includes realization and teing of an (in the final stage) platform-dependent web applicatic Students learn the necessary activities and practice the underlying techniques and tools. The tasks typically relate to practicing the techniques and technologies explained. Literature provided in class Requirements • Proven programming skills in at least one programming language • Knowledge of Software Engineering as thought, for exampling the course Software Engineering. Exam Final presentation, prototype and written elaboration/documentation. Recommended Term MSc 1-3 Turn every winter term Duration 1 Semester Use • MSc AInf Variant: Data Analytics (fundamental trades of the course of the c	, , ,	Students acquire analytical and methodological competencies in the area of modern web technologies and architectures. They recognize and understand the practical constraints relevant to the implementation of modern web applications. They are able to reflect on possibilities and application areas of web technologies as well as associated development environments and to assess them in context. They acquire the technical competences for the realisation of an exemplary web application, especially by working in a team. Through teamwork and exchange with the lecturers, they acquire social skills such as conflict resolution strategies, communication skills, technology and effectiveness assessment.
Requirements • Proven programming skills in at least one programming laguage • Knowledge of Software Engineering as thought, for exampling the course Software Engineering. Exam Final presentation, prototype and written elaboration/documentation. Recommended Term MSc 1-3 Turn every winter term Duration 1 Semester Use • MSc AInf Variant: Data Analytics (fundamental trace—Elective Modules—Application—Computer Science of Software Engineering as thought, for exampling the course of Software Engineering as the course of Software Engineering as the course of Software Engi	Content	In this course, students learn software development for current web technologies. For this purpose, the technical basics of website creation, current web technologies and frameworks, associated development environments and techniques are explained and practiced in the form of an accompanying development project. The development work is done in a team and includes realization and testing of an (in the final stage) platform-dependent web application. Students learn the necessary activities and practice the underlying techniques and tools. The tasks typically relate to practicing the techniques and technologies explained.
 Proven programming skills in at least one programming language Knowledge of Software Engineering as thought, for exampling in the course Software Engineering. Exam Final presentation, prototype and written elaboration/documentation. Recommended Term MSc 1-3 Turn every winter term Duration 1 Semester Use MSc AInf Variant: Data Analytics (fundamental traces and traces are supported by the property of the property of	Literature	provided in class
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ence/Software Engineering	Use	• MSc AInf Variant: Data Analytics (fundamental track) - Elective Modules - Application - Computer Science/Software Engineering
Language English	Language	English

Module: Processes and Management of Software Engineering

Responsible	Prof. Dr. Klaus Schmid
Responsible Instructors	Prof. Dr. Michael Hißmann; Prof. Dr. Klaus Schmid and mem-
_	bers of the study group
Type	2 HPW lecture, 2 HPW tutorial
Credit Points	6 CPs
Workload	presence: 42 hours; self-study: 108 hours
Learning goals/ Compe-	This module teaches methodological and analytical competencies
tencies	that are necessary for a scientific study of processes and management activities in the field of software engineering. In particular, competencies for empirical research in the field of software engineering are taught. This provides students with the necessary expertise to analyze the suitability of software development processes and quality management methods and to develop improvements. The students understand current research questions and approaches.
Content	In this module, the fundamentals of the process models and management activities of software engineering are taught. Special emphasis is placed on the empirical scientific methods of software engineering. In particular, the following topics are addressed:
	 Process modeling and process description languages. Basics of project management (cost estimation, project control) Maturity models and assessments (CMMI, ISO 9000,) Measurement and evaluation (e.g., Goal-Question-Metric) Organizational improvement approaches (QIP, TQM) Configuration management Quality Management
	In the exercise, the contents taught in the lecture: Processes and Management of Software Engineering are deepened by means of exercises. Tasks are worked on together in the exercise and questions concerning the analysis of the methods are discussed. The focus is on imparting the competence for independent analysis and further development by the students.
Literature	Announced in class
Requirements	Knowledge of Software Engineering as thought, for example, in the course Software Engineering.
Exam	written exam (120 min) or an oral exam (30 min)
Recommended Term	MSc 1-3
Turn	every winter term
Duration	1 Semester
Use	• MSc AInf Variant: Data Analytics (fundamental track) - Elective Modules - Application - Computer Science/Software Engineering
Language	English

Module: Software Product Line Engineering

Responsible	Prof. Dr. Klaus Schmid
Responsible Instructors	Prof. Dr. Klaus Schmid and members of the study group
Type	2 HPW lecture, 2 HPW tutorial
Credit Points	6 CPs
Workload	presence: 42 hours; self-study: 108 hours
Learning goals/ Compe-	Students understand the essential differences between single sys-
tencies	tem and product line development, to describe the necessary methodological differences of a product line development and to reflect these in the context of given use cases. They know the current state of science in this field and are able to relate different approaches to each other or to distinguish them from each other. They know the current state of science in this field and are able to classify recent work. They are able to continuously develop their level of knowledge.
Content	The entire software life cycle from the perspective of product line engineering (PLE) is presented. Current approaches from these areas are discussed. Since all sub-activities of software development are affected by PLE, all activities are also examined with respect to changes in a product line approach. Major focal points of the lecture are: 1. Product portfolio planning from a technical perspective and from a market perspective 2. Modeling of variability (decision modeling, feature modeling) 3. Architecture patterns for representing variability 4. Implementation mechanisms for the realization of variability 5. Test strategies 6. Maturity and adaptation models for product line development
	Within the scope of the exercise, the contents taught in the lecture: Software Product Line Development are deepened on the basis of exercises. Tasks are worked on together in the exercise and questions concerning the analysis of the methods are discussed. The focus is on imparting the competence for independent analysis and further development by the students.
Submodules	SM 1: Software Product Line Engineering, Lecture Type: 2 HPW Lecture (3 CPs) Lecturer: Prof. Dr. Klaus Schmid SM 2: Software Product Line Engineering, Tutorium Type: 2 HPW Tutorium (3 CPs) Lecturer: Prof. Dr. Klaus Schmid and members of the study group

Literature	 Frank van der Linden, Klaus Schmid, Eelco Rommes. Product Lines in Action. Springer, 2007. Linda Northrop, Paul Clements. Software Product Lines: Practices and Patterns. Addison-Wesley, 2001. Sven Apel, Don Batory, Christian Kästner, Gunter Saake. Feature-Oriented Software Product Lines: Concepts and Implementation. Springer, 2013. Current scientific publications. Announced in class. Additional Material, made available in class.
Requirements	Knowledge of Software Engineering as thought, for example, in the course Software Engineering.
Exam	written exam (120 min) or an oral exam (30 min)
Recommended Term	MSc 1-3
Turn	every summer term
Duration	1 Semester
Use	• MSc AInf Variant: Data Analytics (fundamental track) – Elective Modules – Application – Computer Science/Software Engineering
Language	English

Module: Software Testing

Responsible	Prof. Dr. Klaus Schmid
Responsible Instructors	Dirk Herrmann
Type	2 HPW lecture (with integrated tutorial)
Credit Points	3 CPs
Workload	presence: 21 hours; self-study: 54 hours
Learning goals/ Competencies	In this module, students acquire detailed knowledge of the field of software testing. They learn the basic procedures of software testing and acquire the necessary practical knowledge to perform this manually and automatically. They will also learn advanced testing methodologies that lead to the optimization of the defect detection rate. They are able to apply these techniques systematically and in a goal-oriented manner.
Content	This course broadly covers the fundamentals of software testing. This includes in particular:
	 Basics of testing (incl. test levels) Black-box/white-box testing Derivation procedures for test cases (boundary cases) Standards Test metrics GUI testing
Literature	Announced in class
Requirements	 Knowledge of Software Engineering as thought, for example, in the course Software Engineering. Good programming knowledge, ideally Java.
Exam	written exam (60 - 90 min) or oral exam (20 min)
Recommended Term	MSc 1-3
Turn	every summer term
Duration	1 Semester
Use	• MSc AInf Variant: Data Analytics (fundamental track) – Elective Modules – Application – Computer Science/Software Engineering
Language	English

Module: Analysis of Software Systems

Responsible	Prof. Dr. Klaus Schmid
Responsible Instructors	Prof. Dr. Klaus Schmid and members of the study group
Type	2 HPW Lecture with Tutorium
Credit Points	3 CPs
Workload	presence: 21 hours; self-study: 54 hours
Learning goals/ Compe-	The students understand the range of different software analysis
tencies	techniques, their advantages and limitations. In particular, they
	know the relevant algorithms and their underlying principles. They
	are able to apply those in the context of specific problems and
	implement some major ones. They also understand the major
	difficulties that are relevant to these tasks.
Content	
	• Data-flow analysis
	• Control-flow analysis
	• Type analysis
	• Interprocedural analysis
Literature	Anders Moller and Michael I. Schwartzbach. Static Program Anal-
	ysis. Lecture Notes Aarhus University.
Requirements	
	• Knowledge of Software Engineering as thought, for example,
	in the course Software Engineering.
	Good programming skills
Exam	written exam (60 - 90 min) or oral exam (20 min)
Recommended Term	MSc 3
Turn	every summer term
Duration	1 Semester
Use	
	MSc AInf Variant: Data Analytics (fundamental track)
	- Elective Modules - Application - Computer Sci-
	ence/Software Engineering
Language	English
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Module: Software Verification

Responsible	Prof. Dr. Klaus Schmid
Responsible Instructors	Prof. Dr. Klaus Schmid and members of the study group
Type	2 HPW lecture and tutorial
Credit Points	3 CPs
Workload	presence: 21 hours; self-study: 54 hours
Learning goals/ Competencies	The students understand the range of different software verification techniques, their advantages and limitations. In particular, they know the relevant verification technologies and their underlying principles. They are able to apply those in the context of specific verification problems and use them for specialised situations. They also understand the range of applicability that can be expected per technique.
Content	Various verification techniques and the underlying solver technologies.
	 Hoare-Style verification Symbolic Execution SAT-based Analysis Satisfiability Modulo Theories (SMT) Model Checking
Literature	 Current scientific publications. Announced in class. Additional Material, made available in class.
Requirements	 Knowledge of Software Engineering as thought, for example, in the course Software Engineering. Good programming skills
Exam	written exam (60 - 90 min) or oral exam (20 min)
Recommended Term	MSc 1-3
Turn	every winter term
Duration	1 Semester
Use	• MSc AInf Variant: Data Analytics (fundamental track) - Elective Modules - Application - Computer Science/Software Engineering
Language	English

Module: Software Architectures

Responsible Instructors Prof. Dr. Klaus Schmid and members of the study group 2 HPW lecture, 2 HPW tutorial Credit Points 6 CPs Workload Learning goals/ Competencies Students learn the foundations of creating, evaluating and realised software architectures. They understand the foundations of architectures and they acquire the competence to define soft architectures for specific systems. They also achieve the capity to evaluate and criticize existing architectures. The student understand the importance of software architectures in the ware development lifecycle and how software architectures realised to business models on the one hand and technical aspects or other hand. Content The course will cover both in a theoretically advanced and a particular tically concrete way the following areas: Principles of good architectures Modeling architectures Architectural styles, patterns, tactics	good tware pabil- dents soft- relate
Credit Points 6 CPs	good tware pabil- dents soft- relate
Workload presence: 42 hours; self-study: 108 hours Learning goals/ Competencies Students learn the foundations of creating, evaluating and realise software architectures. They understand the foundations of architectures and they acquire the competence to define soft architectures for specific systems. They also achieve the capity to evaluate and criticize existing architectures. The student understand the importance of software architectures in the ware development lifecycle and how software architectures realised to business models on the one hand and technical aspects on other hand. Content The course will cover both in a theoretically advanced and a partically concrete way the following areas: Principles of good architectures Modeling architectures Architectural styles, patterns, tactics	good tware pabil- dents soft- relate
Learning goals/ Competencies Students learn the foundations of creating, evaluating and real software architectures. They understand the foundations of architectures and they acquire the competence to define soft architectures for specific systems. They also achieve the cap ity to evaluate and criticize existing architectures. The student understand the importance of software architectures in the ware development lifecycle and how software architectures reached to business models on the one hand and technical aspects on other hand. Content The course will cover both in a theoretically advanced and a particular concrete way the following areas: Principles of good architectures Modeling architectures Architectural styles, patterns, tactics	good tware pabil- dents soft- relate
tencies software architectures. They understand the foundations of a architectures and they acquire the competence to define soft architectures for specific systems. They also achieve the cap ity to evaluate and criticize existing architectures. The student understand the importance of software architectures in the ware development lifecycle and how software architectures reported to business models on the one hand and technical aspects on other hand. Content The course will cover both in a theoretically advanced and a partically concrete way the following areas: • Principles of good architectures • Modeling architectures • Architectural styles, patterns, tactics	good tware pabil- dents soft- relate
tically concrete way the following areas: • Principles of good architectures • Modeling architectures • Architectural styles, patterns, tactics	
 Modeling architectures Architectural styles, patterns, tactics	prac-
 Design approaches Architecture evaluation Modern architecture paradigms like service-orientation Architectures for Big Data Systems 	
Submodules SM 1: Software Architectures, Lecture Type: 2 HPW Lecture CPs) Lecturer: Prof. Dr. Klaus Schmid SM 2: Software Architectures, Tutorium Type: 2 HPW Tutorium (3 CPs) Lecturer: Dr. Klaus Schmid and members of the study group	hitec-
Literature	
 K. Bass, P. Clements, R. Kazman: Software architectw practice. Addison-Wesley, 2012. R.Kazman, H. Cervantes: Designing Software Architect Addison-Wesley, 2016. Current scientific publications. Announced in class. Additional Material, made available in class. 	
Requirements	
 Knowledge of Software Engineering as thought, for examine the course Software Engineering. Good Programming Skills, preferably in Java, but other guages are also possible 	
Exam written exam (120 min) or an oral exam (30 min)	
Recommended Term MSc 1-3	
Turn every winter term	

Use	
	• MSc AInf Variant: Data Analytics (fundamental track) - Elective Modules - Application - Computer Science/Software Engineering
Language	English

Module: Requirements Engineering for Quality

Responsible	Prof. Dr. Klaus Schmid
Responsible Instructors	Prof. Dr. Michael Hißmann; Prof. Dr. Klaus Schmid and mem-
	bers of the study group
Type	2 HPW lecture, 2 HPW tutorial
Credit Points	6 CPs
Workload	presence: 42 hours; self-study: 108 hours
Learning goals/ Competencies	Students acquire the essential methodological and theoretical basics in requirements engineering. They can use different methods within the context of specific development situations and they are able to reflect the limits and possibilities of different approaches. Students can indepently adapt these approaches to specific contexts.
Content	This lecture introduces methodological and theoretical basics of requirements engineering. The components of requirements engineering and state-of-the-art techniques and their application are discussed In particular:
	 elicitation techniques requirements analysis and modelling creating contracts and requirements specifications quality requirements (reliability, security, usability)
	During the tutorium students deepen their knowledge of the material taught in the lectures. They solve tasks together and do homeworks. The tutorium focusses on communication, problem-solving-competence and transfer-competence.
Literature	 C. Rupp: Requirements Engineering und Management. Hanser, 2020. K. Pohl: Requirements Engineering: Fundamentals, Principles, and Techniques. Springer, 2016. Current scientific publications. Announced in class. Additional Material, made available in class.
Requirements	Knowledge of Software Engineering as thought, for example, in the course Software Engineering.
Exam	written exam (120 min) or an oral exam (30 min)
Recommended Term	MSc 1-3
Turn	every winter term
Duration	1 Semester
Use	• MSc AInf Variant: Data Analytics (fundamental track) – Elective Modules – Application – Computer Science/Software Engineering
Language	English

Module: SAT Solving

Responsible	Dr. Jean Christoph Jung
Lecturer	Dr. Jean Christoph Jung
Туре	2 HPW Seminar
Credit Points	4 CPs
Workload	presence: 21 hours; self-study: 79 hours
Learning goals/ Competencies	The goal of the seminar is the independent reading and understanding of original scientific literature, the composition of a scientific report on the selected paper, and finally the presentation of the material to the other participants of the seminar. This will enable the students to acquire the basic skills for the scientific
	method.
Content	The seminar will deal with automated reasoning mostly for satisfiability (SAT) problems. These reasoning procedures were coined "SAT solvers" and are one success story of the field of artificial intelligence: In spite of solving an NP-complete (and thus "intractable") problem, SAT solvers perform quite well in practice due to sophisticated implementation techniques. We will cover the basic techniques in the seminar: clause learning, watched literal schemes and unit propagation, local search, parallelization, variable and value selections, and more. Possible other aspects are verification, satisfiability modulo theories (SMT), modeling, and counting solutions to SAT problems.
Submodules	keine
Literature	• Biere, A., Heule, M., van Maaren, H. and Walsh, T. (Eds.): Handbook of Satisfiability, Frontiers in Artificial Intelligence and Applications, 2009.
Requirements	keine
Exam	Ausarbeitung und Vortrag
Recommended Term	BSc 4-6 oder MSc 1-3
Turn	unregelmäßig
Duration	1 Semester
Use	• MSc AInf Variant: Data Analytics (fundamental track) – Elective Modules – Application – Computer Science/Software Engineering
Language	English

Module: Efficient Algorithms

Modulverantwortlicher	Dr. Jean Christoph Jung
Lehrform/SWS	2 SWS Seminar
Leistungspunkte	4 LP
Lernziele/Kompetenzen	The goal of the seminar is the independent reading and understanding of original scientific literature, the composition of a scientific report on the selected paper, and finally the presentation of the material to the other participants of the seminar. This will enable the students to acquire the basic skills for the scientific method.
Lehrinhalte	The seminar will deal with the design and analysis of efficient algorithms for different problems that are typically not covered in the basic module 'Einführung in die Informatik 2.'
Teilmodule / Veranstaltungen	keine
Literatur	Cormen, Leiserson, Rivest, Stein: Introduction to Algorithms, MIT Press 2009
Voraussetzungen für die	Einführung in die Informatik 2 (Algorithmen und Datenstruk-
Teilnahme	turen)
Prüfungsleistung	Ausarbeitung und Vortrag
empfohlenes Semester	BSc 4-6 oder MSc 1-3
Turnus	unregelmäßig
Dauer des Moduls	1 Semester
Verwendung	 Informationsmanagement und Informationstechnologie (IMIT) / BSc. und MSc. Wahlmodul Informatik – Gebiet Theoretische Informatik Angewandte Informatik / BSc. und MSc. Wahlmodul Informatik – Gebiet Theoretische Informatik Wirtschaftsinformatik BSc. und MSc. – Wissensmanagment Master Programme Data Analytics – Elective Modules – Application – Computer Science / Software Engineering
Use	• MSc AInf Variant: Data Analytics (fundamental track) – Elective Modules – Application – Computer Science/Software Engineering

Module: SE Tools Lab

Responsible	Prof. Dr. Klaus Schmid
Responsible Instructors	Prof. Dr. Klaus Schmid and members of the study group
Type	4 HPW lab course
Credit Points	6 CPs
Workload	presence: 42 hours; self-study: 108 hours
Learning goals/ Compe-	Students learn about different classes of software development tools
tencies	that cover the essential phases of software development. Subse-
	quently, the students have the necessary competences for the in-
	dependent selection and application of tools.
Content	In this lab course, students learn the use of various software devel-
	opment tools and techniques. This includes tools for requirements
	engineering, modeling of software systems, version management
	and continuous integration. The lab course complements the con-
	tents of the Software Engineering course.
Literature	material handed out in class.
Requirements	Knowledge of Software Engineering as thought, for example, in the
	course Software Engineering.
Exam	multiple tasks with oral examination
Recommended Term	MSc 1-3
Turn	every summer term
Duration	1 Semester
Use	
	MSc AInf Variant: Data Analytics (fundamental track)
	- Elective Modules - Application - Computer Sci-
	ence/Software Engineering
Language	English

Module: Master-Seminar Software Engineering

Type	2 HPW seminar
Credit Points	4 CPs
Workload	presence: 21 hours; self-study: 79 hours
Learning goals/ Compe-	The objective of this seminar is the autonomous exploitation and
tencies	elaboration of a predetermined subject. The participation in
	course and the scientific discussion of the lectures will deepen the
	understanding of the knowledge already acquired. Acquirement
	of methodological competence: students gain transfer competence
	allowing them to autonomously adjust their level of knowledge to
	technical and social development.
Content	Different advanced topics from the field of modern Software Engi-
	neering.
Submodules	none
Literature	Depending on the topic
Requirements	Basic knowledge of Software Engineering.
Exam	Seminar paper and oral presentation
Recommended Term	MSc 2-3
Turn	every semester
Duration	1 Semester
Responsible	Prof. Dr. Klaus Schmid
Use	
	• MSc AInf Variant: Data Analytics (fundamental track)
	– Elective Modules – Application – Computer Sci-
	ence/Software Engineering

Business Administration and Information Systems

Module: Advanced Marketing

Responsible	Prof. Dr. Julia Rieck
Responsible Instructors	Prof. Dr. Julia Rieck and members of the study group
Type	3 HPW lecture
Credit Points	4 CPs
Workload	presence: 32 hours; self-study: 68 hours
Learning goals/ Compe-	The students know different analysis methods for marketing data
tencies	and use them specifically. Furthermore they can analyse market research data with the use of appropriate evaluation programs. Students are able to distinguish between alternative methods and to recognize possible limits. Furthermore they are capable to analyse and evaluate data sets and interpret the results. They are able to understand and to implement current developments independently.
Content	Topics covered on the course include market research relevant
	methods of data collection and analysis (supervised and unsuper-
	vised methods) and application in marketing.
Submodules	none
Literature	 Malhotra, N., Nunan, D., Birks, D.: Marketing Research: An applied approach, Pearson Education Limited, 5th Edition, 2017 Kotler, P., Armstrong, G., Harris, L., Piercy, N.: Principles of Marketing (European Edition), Pearson, 7th Edition, 2016 Sorger, S.: Marketing Analytics: Strategic Models and Metrics, CreateSpace Independent Publishing Platform, 2013
Requirements	Machine Learning
Exam	written exam (120 min)
Term	MSc 1-3
Turn	every summer term
Duration	1 Semester
Use	• MSc AInf Variant: Data Analytics (fundamental track) – Elective Modules – Application – Business Administration and Information Systems
Language	English

Module: Product development and technologies for navigation and driver assistance systems

Responsible	Prof. Dr. Julia Rieck, Dr. Felix Hahne
Responsible Instructors	Dr. Thomas Kleine-Besten and others
Type	2 HPW lecture
Credit Points	3 CPs
Workload	presence: 21 hours; self-study: 54 hours
Learning goals/ Competencies	The students are acquainted with the tasks and challenges of a product development on the basis of real examples. As example products navigation and driver assistance systems are used, which are developed in the automotive development process. In addition the business aspects, the students are familiar with the underlying technologies as well. The students know the basic tasks of a product development of a complex technical product in the business environment as well as its interaction with the technical conditions. They can classify the learned content in the context of the scientific discipline and connect it to the knowledge learned so far in business economics. A discussion of the topics covered take
Content	Students are introduced to the product development process in the automotive-industry and learn about the underlying technologies using navigation- und driver assistance systems as an example. Topics covered are e.g. • Marketing, product management • Commercial acquisition process • Technical customer acquisition: hardware and software platforms • Requirements analysis and automotive development process • Project management • Introduction to navigation systems • Bluetooth • Driver assistance • Application: The "electronic horizon" • Car to Car – Communications • Machine Learning • Digital Maps for highly-automated driving • Testing procedures
Submodules	none

Literature	
	 Winner, H., Hakuli, S., Lotz, F., Singer, C. (2015): Handbuch Fahrerassistenzsysteme: Grundlagen, Komponenten und Systeme für aktive Sicherheit und Komfort, 3. Auflage, Springer Vieweg, Wiesbaden Schäuffele, J., Zurawka, T. (2016): Automotive Software Engineering: Grundlagen, Prozesse, Methoden und Werkzeuge effizient einsetzen, 6. Auflage, Springer Vieweg, Wiesbaden Rupp C., die SOPHISTen (2009): Requirements-Engineering und Management: Professionelle, iterative Anforderungsanalyse für die Praxis, 5. Auflage, Hanser, München Krüger, R. (2004): Lehr- und Übungsbuch Telematik, 3. Auflage, Hanser, München Merkle, A., Terzis, A. (2002): Digitale Funkkommunikation mit Bluetooth, Franzis, Haar Mulcahy, R., PMP Exam Prep (2013): Rita's Cource in a Book for Passing the Pmp E, Bertrams, Hilden
Requirements	none
Exam	written exam (90 - 120 min). Possibly additional requierements have to be met to be admitted to the final exam - further information is available in the accompanying learnweb-course.
Term	MSc 1-3
Turn	Each summer term
Duration	1 Semester
Use	
	• MSc AInf Variant: Data Analytics (fundamental track) – Elective Modules – Application – Business Administration and Information Systems
Language	English

Module: Project Management and Scheduling

Responsible	Prof. Dr. Julia Rieck
Responsible Instructors	Prof. Dr. Julia Rieck
Type	2 HPW lecture, 2 HPW tutorial
Credit Points	6 CPs
Workload	presence: 42 hours; self-study: 108 hours
Learning goals/ Compe-	Upon completion of this course, the students can:
tencies	 fully understand fundamental scheduling and sequencing problems that arise in resource-constrained project scheduling environments within the manufacturing and service industry, apply state-of-the-art methodologies for effectively and efficiently planning projects subject to both precedence and resource constraints, manage and control a project.
Content	Project representation using activity networks, time analysis (estimating the project duration in a deterministic setting), resource management, i.e. resource leveling (leveling the use of the resources over time subject to a project deadline) and resource-constrained-project scheduling (scheduling the activities subject to the various precedence and resource constraints in order to minimize the project duration and other objective functions).
Submodules	none
Literature	 Neumann, K.; Schwindt, C.; Zimmermann, J. (2003): Project Scheduling with Time Windows and Scarce Resources, 2nd edition, Springer, Berlin Schwindt, C.; Zimmermann, J. (2015): Handbook on Project Management and Scheduling Vol. 1, Springer, Cham Schwindt, C.; Zimmermann, J. (2015): Handbook on Project Management and Scheduling Vol. 2, Springer, Cham Vanhoucke, M. (2013): Project Management with Dynamic Scheduling: Baseline Scheduling, Risk Analysis and Project Control, 2nd edition, Springer, Berlin
Requirements	none
Exam	Written exam (90 min); for justified exceptions oral exam (30 min). Possibly additional requierements have to be met to be admitted to the final exam - further information is available in the accompanying learnweb-course.
Term	MSc 1-3
Turn	Each winter term
Duration	1 Semester

Use	• MSc AInf Variant: Data Analytics (fundamental track) – Elective Modules – Application – Business Administration and Information Systems
Language	English

Module: Seminar Business Studies

Responsible	Prof. Dr. Julia Rieck, Dr. Felix Hahne
Responsible Instructors	Prof. Dr. Julia Rieck, Dr. Felix Hahne and members of the study
	group
Type	2 HPW seminar
Credit Points	4 CPs
Workload	presence: 21 hours; self-study: 79 hours
Learning goals/ Competencies	The objective of this seminar is the autonomous exploitation and elaboration of a predetermined subject. The participation in course and the scholarly debate about the presentations are to lead to a deeper understanding of previously acquired knowledge. Acquirement of methodological competence: students gain transfer competence allowing them to autonomously adjust their level of knowledge to technical and social development. Depending on the area of specialization, students acquire different economical competences and become acquainted with different instruments. Possible areas of specialization include: marketing, logistics, production, business intelligence systems.
Content	Different advanced topics from the fields of marketing (e.g. marketing research, marketing policy), logistics (e.g. transportation planning, location planning, warehousing), production, and business intelligence systems (e.g. support of business functions, opening of new business areas).
Submodules	none
Literature	depending on the topic
Requirements	The contents of a correspondent master's course in the selected topic are implied (e.g. Advanced Marketing, Advanced Logistics, Advanced Production, Business Intelligence Systems)
Exam	Seminar paper and oral presentation
Recommended Term	MSc 2-3
Turn	every winter term
Duration	1 Semester
Use	• MSc AInf Variant: Data Analytics (fundamental track) – Elective Modules – Application – Business Administration and Information Systems
Language	English

Module: Business Intelligence and Data Warehousing

Responsible	Prof. Dr. Lars Schmidt-Thieme
Responsible Instructors	Prof. Dr. Lars Schmidt-Thieme
Type	2 HPW lecture, 2 HPW tutorial
Credit Points	6 CPs
Workload	presence: 42 hours; self-study: 108 hours
Learning goals/ Competencies	Business Intelligence seeks to extract and present insights from operational data that are relevant to future decisions. In larger companies, it is common practice to provide the operating data for this in one place data warehouse adjusted and systematized
Content	This lecture deals with the basics of the task, the construction and realization of data warehouses and the embedding of data warehousing in the overall context of Business Intelligence.
Literature	
	 J. Celko: Joe Celko's Data Warehouse and Analytic Queries in SQL, (2006) ISBN-13: 978-0123695123 Graziano, Linstedt: Super Charge Your Data Warehouse, (2011) ISBN-13: 978-1463778682 W.H. Inmon: Building the Data Warehouse, (2005) ISBN-13: 978-0764599446 J.E.Olson: Data Quality: The Accuracy Dimension, (2002) ISBN-13: 978-1558608917
Requirements	none
Exam	written exam (120 min) or an oral exam (30 min)
Recommended Term	MSc 1-3
Turn	every second term
Duration	1 Semester
Use Language	MSc AInf Variant: Data Analytics (fundamental track) – Elective Modules – Application – Business Administration and Information Systems English
Language	- Duguen

Module: Data Warehousing in Practice

Responsible	Prof. Dr. Lars Schmidt-Thieme
Responsible Instructors	Christoph Seck
Type	2 HPW lecture, 2 HPW tutorial
Credit Points	6 CPs
Workload	presence: 42 hours; self-study: 108 hours
Learning goals/ Compe-	Business Intelligence seeks to extract and present insights from
tencies	operational data that are relevant to future decisions. In larger
	companies, it is common practice to have the operational data in
	one place "data warehouse" cleared up and systematized.
Content	The core of the lecture is the development of a data warehouse
	based on a concrete example. The starting point is a seemingly
	simple problem from a manufacturing company. Even more than
	in Business Intelligence and Data Warehousing 1 (eng.), exercises
	and practical work will be at the center. The focus will, therefore,
	be on the aspects of the DWH process that take up most of the
	space in such projects: analysis of source systems, handling of
	insufficient requirements, data quality problems, unusual reporting requirements, etc. In addition, the lecture will deal with a few
	topics that were not dealt with in the last semester, or only in
	passing. These include Master Data Management, Data Vault and
	the current hypertexts Self Service BI and Big Data
Submodules	SM 1: Data Warehousing in Practice, Lecture Type: 2 HPW Lec-
	ture (3 CPs) Lecturer: Christoph Seck SM 2: Data Warehousing
	in Practice, Tutorium Type: 2 HPW Tutorium (3 CPs) Lecturer:
	Christoph Seck
Literature	
	• J. Celko: Joe Celko's Data Warehouse and Analytic Queries
	in SQL, (2006) ISBN-13: 978-0123695123
	• Graziano, Linstedt: Super Charge Your Data Warehouse,
	(2011) ISBN-13: 978-1463778682
	• W.H. Inmon: Building the Data Warehouse, (2005) ISBN-13:
	978-0764599446
	• Kimball, Ross: The Data Warehouse Toolkit, (2013) ISBN-
	13: 978-1118530801 • Kimball, Munday, Thronthwaite: The Microsoft Data Ware-
	house Toolkit, (2011) ISBN-13: 978-0470640388
	• J.E.Olson: Data Quality: The Accuracy Dimension, (2002)
	ISBN-13: 978-1558608917
	• Russo, Ferrari, Webb: Expert Cube Development with Mi-
	crosoft SQL Server 2008 Analysis Services, (2009) ISBN-13:
	978-1847197221
	• Russo, Ferrari, Webb: Microsoft SQL Server 2012 Analysis
	Services: The BISM Tabular Model, (2012) ISBN-13: 978-
	0735658189
	• NBI Testing Tool und Dokumentation auf: http://nbi.
	codeplex.com/

Requirements	none
Exam	written exam (120 min) or an oral exam (30 min)
Recommended Term	MSc 1-3
Turn	every winter term
Duration	1 Semester
Use	• MSc AInf Variant: Data Analytics (fundamental track) – Elective Modules – Application – Business Administration and Information Systems
Language	English

Information Retrieval and Information Sciences

Module: Introduction Information Retrieval (IR)

Responsible	Prof. Dr. Thomas Mandl
Lecturer	Prof. Dr. Thomas Mandl and members of the study group
Type	2 HPW lecture
Credit Points	4 CPs
Workload	presence: 21 hours; self-study: 79 hours
Learning goals/ Competencies	Students know the technologies for representing Information Retrieval Systems and are familiar with search models. They should be able to describe Information Retrieval Systems and their components as well as assign them to fundamental paradigms. The Students are able to differentiate Information Retrieval systems from the area of Databases. They know how to use user-oriented processes for evaluating Information Processes.
Content	Information Retrieval deals with uncertain representation of unstructured knowledge (especially text) and a vague search of information. The lecture gives an overview of Retrieval-Processes and introduces in detail manual and automatic indexing as well as weighting and treats important search models (partial and exact match, vector space, language model). One main focuss are evaluation approaches. Other contents are user behavior, user interface, web-retrieval and multimedia-retrieval.
Literature	• R. Manning, H. Schütze: Introduction to Information Retrieval Cambridge University Press. 2008.
Requirements	none
Exam	written exam (60 - 90 min) or oral exam (20 min)
Recommended Term	MSc 1-3
Turn	every year
Duration	1 Semester
Use	• MSc AInf Variant: Data Analytics (fundamental track) – Elective Modules – Application – Information Retrieval and Information Sciences
Language	English

Module: Introduction Natural Language Processing

Responsible	Prof. Dr. Ulrich Heid
Lecturer	Prof. Dr. Ulrich Heid and members of the study group
Type	3 HPW lecture
Credit Points	4 CPs
Workload	presence: 32 hours; self-study: 68 hours
Learning goals/ Compe-	The Students know the most important symbolic and statistical
tencies	language process methods and are able to estimate their performance and limits as well as their application relevance. They are able to understand and give a professional opinion on model and implementation approaches. The students know evaluation approaches for language processing systems and are able to evaluate on there own. During the tutorium they gain knowledge how to install and use language processing tools and know their functunality, their Input and Output and their resource requirements. They are able to interpret the different outcomes and judge them related to the explicit task. Some example of those tools are Tokenizer, Wordclass-Tagging, morphological and syntactical analysis systems (Parser) and so on.
Content	Tasks, methods, processes and application of language processing. The main focuss are the fundamental rule-based and statistic procedures for automatic processing of written language; especially the ones who are important for information science applications (e.g. Information retrieval, Imformation-Extraction, Multilingual Applications). Evaluation methods and principles. The tutorium starts the third week of the semester. First an introduction in Linux is given. Since week 3: exercises depending on the topics of the lecture: procedures, methods and application of language processing. The focuss lies on the practical use of language processing tools available from the institute or the internet.
Submodules	SM 1: Introduction Natural Language Processing, Lecture Type: 3 HPW Lecture (4 CPs)
Literature	will be announced in the lecture
Requirements	basic knowledge in Information Systems.
Exam	written exam (60 - 90 min) or oral exam (20 min)
Recommended Term	MSc 1-3
Turn	every year
Duration	1 Semester
Use	MSc AInf Variant: Data Analytics (fundamental track) – Elective Modules – Application – Information Retrieval and Information Sciences
Language	English

Module: Multilingual Information Systems

Responsible	Prof. Dr. Thomas Mandl
Lecturer	Prof. Dr. Thomas Mandl and members of the study group
Type	2 HPW lecture
Credit Points	4 CPs
Workload	presence: 21 hours; self-study: 79 hours
Learning goals/ Compe-	The students should have developed a deeper understanding in
tencies	multilingual information systems. They are able to use such sys-
	tems target-oriented and apply evaluation methods to multilingual
	information systems.
Content	The multilingual content in Information Systems increases and
	must be handled. For example Information Retrieval or Text Mining to multiple language, structure and management of multilin-
	gual knowledge bases, software-localization as well as databases
	with multilingual content. Content of this course are methods,
	systems, evaluation methods and problems with the usage of in-
	formation systems in multilingual areas. The students deepen their
	knowledge in multilingual information systems. They are able to
	use such systems target-oriented and apply evaluation methods to
	multilingual information systems.
Submodules	SM 1: Introduction Information Retrieval (IR), Lecture Type: 2
	HPW Lecture (4 CPs)
Literature	will be announced in the lecture
Requirements	Contents of the lecture "Introduction Information Retrieval (IR)".
Exam	written exam (60 - 90 min) or oral exam (20 min)
Recommended Term	MSc 1-3
Turn	every year
Duration	1 Semester
Use	
	• MSc AInf Variant: Data Analytics (fundamental track) –
	Elective Modules – Application – Information Retrieval and
	Information Sciences
Language	English

Module: Seminar Multilingiual Information Retrieval

Responsible	Prof. Dr. Thomas Mandl
Responsible Instructors	Prof. Dr. Thomas Mandl and members of the study group
Type	2 HPW seminar
Credit Points	4 CPs
Workload	presence: 21 hours; self-study: 79 hours
Learning goals/ Compe-	The students are familiar with Multilingual Inforamtion Retrieval
tencies	Systems, know the problems, tools and user-oriented evaluation
	methods. They are able to exploit and elaborate a predetermined
	problem.
Content	This Seminar deepens the knowledge of Multilingual Informa-
	tion Retrieval, presents state-of-the-art language-dependent and
	language-independent methods and shows tools for Multilingual
	Retrieval.
Submodules	none
Literature	depending on the topic
Requirements	The contents of the lecture "Multilingual Information Systems"
Exam	Active participation and written seminar paper.
Recommended Term	MSc 2-3
Turn	every year
Duration	1 Semester
Use	
	• MSc AInf Variant: Data Analytics (fundamental track) –
	Elective Modules – Application – Information Retrieval and
	Information Sciences
Language	English

Module: Project Multilingual Information Systems

Dagnangible	Prof. Dr. Thomas Mandl
Responsible	
Responsible Instructors	Prof. Dr. Thomas Mandl and members of the study group
Type	4 HPW project
Credit Points	6 CPs
Workload	presence: 42 hours; self-study: 108 hours
Learning goals/ Compe-	In this project students have the possibility to choose suitable
tencies	methods and use them target-oriented. Reflected and led by theory
	students strive for a praxisoriented solution. The students should
	be able to organize and structure their project in a small team.
Content	In this course students perform a specific small-scale-project based
	on state-of-the-art researches in Multilingual Information Systems.
Submodules	none
Literature	depending on the topic
Requirements	The contents of the Module "Multilingual Information Systems"
Exam	written summary
Recommended Term	MSc 3
Turn	every year
Duration	1 Semester
Use	
	• MSc AInf Variant: Data Analytics (fundamental track) –
	Elective Modules – Application – Information Retrieval and
	Information Sciences
Language	English

Module: Lab Course Information Retrieval (IR)

Responsible	Prof. Dr. Thomas Mandl
Lecturer	Prof. Dr. Thomas Mandl and members of the study group
Type	2 HPW lab course
Credit Points	4 CPs
Workload	presence: 21 hours; self-study: 79 hours
Learning goals/ Compe-	Students are able to use tools for every phase in Information Re-
tencies	trieval Processes. They can use systems target- and task-oriented
	and evaluate them depending on the situation.
Content	The Praktikum focusses on tools for Information Retrieval and
	their components:
	 manual indexing based on a classification system automatic indexing (stemming) and its evaluation search methods and search tools relevance-feedback and term extension relevance-evaluation and evaluation methods
Submodules	none
Literature	• R. Manning, H. Schütze: Introduction to Information Retrieval Cambridge University Press. 2008.
Requirements	Contents of the lecture "Introduction Information Retrieval (IR)".
Exam	Homework and written exam.
Recommended Term	MSc 2-3
Turn	every year
Duration	1 Semester
Use	
	• MSc AInf Variant: Data Analytics (fundamental track) – Elective Modules – Application – Information Retrieval and Information Sciences
Language	English

Natural Language Processing

Module: Natural Language Processing

Responsible	Prof. Dr. Lars Schmidt-Thieme
Lecturer	Prof. Dr. Christian Wartena and members of the study group
Type	2 HPW lecture, 2 HPW tutorial
Credit Points	6 CPs
Workload	presence: 42 hours; self-study: 108 hours
Learning goals/ Competencies	The students have an understanding of structures in natural languages and the traditional symbolic and statistical approaches to model structures in phonology, morphology, syntax and lexical semantics. They have an overview of state-of-the-art Natural Language Processing (NLP) methods. Students are able to build applications using or extending standard NLP methods. They are able to familiarize quickly with other topics in NLP and are able to read original research literature.
Content	Student will learn the most important phenomena in natural languages on different levels of granularity, starting with the combination of sounds to the meaning of words, sentences and texts. You will get an introduction to main symbolic and statistical approaches to model these phenomena. All theoretical topics will be accompanied by exercises dealing with these phenomena and demonstrating their use in practical applications, like spelling correction, auto completion, keyword extraction, topic detection, named entity recognition, relation extraction, synonym detection, etc. Students will apply the basic natural language processing methods in the implementation of a small application or in the analysis of a data set. For this part they will study one specific phenomenon into depth and will be free to explore various machine learning and natural language processing techniques to find their own solution.
Submodules	SM 1: Natural Language Processing 2, Lecture Type: 2 HPW Lecture (2 CPs) Lecturer: Prof. Dr. Christian Wartena SM 2: Natural Language Processing 2, Tutorium Type: 2 HPW Tutorium (2 CPs) Lecturer: Prof. Dr. Christian Wartena and members of the study group
Literature	
	 Bird, Steven, Ewan Klein, and Edward Loper. Natural language processing with Python: analyzing text with the natural language toolkit. O'Reilly Media, Inc. (2009) Daniel Jurafsky and James H. Martin: Speech and Language Processing. Prentice Hall; 2nd edition (2008) Christopher D. Manning and Hinrich Schütze: Foundations of Statistical Natural Language Processing. The MIT Press (1999) Additional literature will be announced in the lecture

Requirements	Basic knowledge of theoretical Computer Science (Automata the-
	ory, complexity, rewriting systems, Markov models).
Exam	Several tests during the semester and/or written exam (60 - 90
	min) or oral exam (20 min) or seminar paper
Recommended Term	MSc 2-3
Turn	irregular
Duration	1 Semester
Use	 MSc AInf Variant: Data Analytics (fundamental track) – Elective Modules – Application MSc AInf Variant: Data Analytics (fundamental track) – Elective Modules – Application – Natural Language Processing
Language	English

Module: Language Modelling

Responsible	Prof. Dr. Ulrich Heid
Lecturer	Prof. Dr. Ulrich Heid and members of the study group
Type	2 HPW lecture 2 HPW project
Credit Points	6 CPs
Workload	presence: 42 hours; self-study: 108 hours
Learning goals/ Compe-	The students should have developed an understanding of struc-
tencies	tures in natural languages and the traditional symbolic and statistical approaches to model structures in phonology, morphology, syntax and lexical semantics. The students should understand the relevance of language modelling for information retrieval, information extraction and natural language processing. The students should be able to implement a model for one isolated linguistic phenomenon and be able to identify potentials of machine learning for that phenomenon.
Content	In the lecture the most important symbolic and statistical models for phonology, morphology and syntax, like e.g. optimality theory, two level morphology, Hidden Markov Models, phrase structure grammars and dependency grammars as well as their application to part-of-speech tagging, lemmatization and parsing will be presented. In the project one topic will be studied into depth using original literature. Students will investigate the possibilities to apply machine learning to improve the model or to reduce the implementation costs. The students will make a prototypical implementation.
Submodules	SM 1: Natural Language Processing 2, Lecture Type: 2 HPW Lecture, 2 HPW Project (6 CPs)
Literature	
	 Daniel Jurafsky and James H. Martin: Speech and Language Processing. Prentice Hall; 2nd edition (2008) Christopher D. Manning and Hinrich Schütze: Foundations of Statistical Natural Language Processing. The MIT Press (1999)
	1. Additional literature will be announced in the lecture
Requirements	Basic knowledge of theoretical computer science (Automata theory, complexity, rewriting systems, Markov models).
Exam	written exam (120 min) or seminar paper
Recommended Term	MSc 2-3
Turn	irregular

Use	• MSc AInf Variant: Data Analytics (fundamental track) – Elective Modules – Application – Natural Language Processing
Language	English

Module: Natural Language Processing 2

Responsible	Prof. Dr. Ulrich Heid
Lecturer	Prof. Dr. Ulrich Heid and members of the study group
Type	2 HPW lecture
Credit Points	4 CPs
Workload	presence: 21 hours; self-study: 79 hours
Learning goals/ Compe-	Students have a great overview about state-of-the-art Natural Lan-
tencies	guage Processing methods. They are able to estimate advantages and limits of those methods; they are able to make transfer relations to aspects of Information Research and Human-Machine-Interaction. The Students are familiar with these method as they can give a professional opinion on their input, resources, output and integration of applications-
Content	The lecture treats selected areas in Natural Language Processing and Language Technology in detail with respect to state-of-the-art research, international as well as research at the institut. It focusses on methods and tools based on theses methods. Some topics are: 1. Analysis and annotation of textdata (tagging, parsing, annotation methods and representations, standards for interoperable annotated corpora, etc.) 2. Methods and paradigms of Language Processing evaluation: evaluation methods, degrees, gold standard, shared tasks, etc. 3. Statistical methods for Language Processing: lexicostatistics, co-occurences analysis, statistical parsing, statistical machine translation, etc. 4. Language Technology as method and tool: digital-humanities applications, Language Processing daily life tools (e.g. dialogue systems, correction of orthography, style-checking, etc.) These topics are treated in a lecture giving an overview and if necessary also from a different focus.
Submodules	SM 1: Natural Language Processing 2, Lecture Type: 2 HPW
	Lecture (4 CPs)
Literature	will be announced in the lecture
Requirements	basic knowledge of Natural Language Processing.
Exam	Several tests during the semester and/or written exam (60 - 90 min) or oral exam (20 min). Homework
Recommended Term	MSc 2-3
Turn	irregular
Duration	1 Semester

Use	• MSc AInf Variant: Data Analytics (fundamental track) – Elective Modules – Application – Natural Language Processing
Language	English

Module: Seminar Computer Linguistic Resources

Responsible	Prof. Dr. Ulrich Heid
Responsible Instructors	Prof. Dr. Ulrich Heid and members of the study group
Type	2 HPW seminar
Credit Points	4 CPs
Workload	presence: 21 hours; self-study: 79 hours
Learning goals/ Compe-	The students have a detailed knowledge about one part in Com-
tencies	puter Linguistic Resources or applications in Language Technology.
	They are able to adapt and evaluate these processes and applica-
	tions to minor research topics. The students can independently
	solve questions of this specific part.
Content	This seminar deepens the knowledge of specific topics in Computer
	Linguistics and Language Technology; Students are led to own mi-
	nor research topics, ideally corresponding to the instituts research.
	Language resources, its establishemnet, management and usage:
	e.g. corpus linguistics, annotation of corpora, building corpus, data
	extraction from corpora; electronic dictionaries, terminology data
	bases, special lexicons for Language Technology (e.g. Sentiment-
	Lexicon); standards for Language resources, Language Resources
	applications, e.g. in the area of digital humanities or iCALL.
Submodules	none
Literature	depending on the topic
Requirements	The contents of the lecture "Natural Language Processing 2"
Exam	Colloquium and written summary.
Recommended Term	MSc 2-3
Turn	irregular
Duration	1 Semester
Use	
	• MSc AInf Variant: Data Analytics (fundamental track) –
	Elective Modules – Application – Natural Language Process-
	ing
Language	English

Module: Seminar Computer Linguistic Processes

Responsible	Prof. Dr. Ulrich Heid
Responsible Instructors	Prof. Dr. Ulrich Heid and members of the study group
Type	2 HPW seminar
Credit Points	4 CPs
Workload	presence: 21 hours; self-study: 79 hours
Learning goals/ Compe-	The students have a detailed knowledge about one part in Com-
tencies	puter Linguistic Processes or applications in Language Technology.
	They are able to adapt and evaluate these processes and applica-
	tions to minor research topics. The students can independently
	solve questions of this specific part.
Content	This seminar deepens the knowledge of specific topics in Computer
	Linguistics and Language Technology; Students are led to own
	minor research topics, ideally corresponding to the instituts re-
	search: Computer Linguistic Processes methods and applications:
	e.g. syntactic-semantic analysis, generation, models of dialogs, ma-
	chine translation.
Submodules	none
Literature	depending on the topic
Requirements	The contents of the lecture "Natural Language Processing 2"
Exam	Colloquium and written summary.
Recommended Term	MSc 3
Turn	irregular
Duration	1 Semester
Use	
	• MSc AInf Variant: Data Analytics (fundamental track) –
	Elective Modules – Application – Natural Language Process-
	ing
Language	English

Module: Project Computer Linguistic Resources

Responsible	Prof. Dr. Ulrich Heid
Responsible Instructors	Prof. Dr. Ulrich Heid and members of the study group
Type	4 HPW project
Credit Points	6 CPs
Workload	presence: 42 hours; self-study: 108 hours
Learning goals/ Compe-	The students are able to analyse and evaluate Computer Linguistic
tencies	Resources. The students should be able to independently solve,
	analyse and implement or respectively adapt or optimize questions
	from the area of Compuer Linguistic Resources. They are able to
	relate their solution to state-of-the-art research. This seminar gives
	the foundations for the master thesis.
Content	Specialization and integrated, theoretical, methodological and
	practical research-based treatment of topics in Computer Linguis-
	tic and Language technology with focus on Resources. The stu-
	dents independently solve the task with Computer Linguistic re-
	sources (if necessary in the case of joint projects like participate
	shared tasks, resource creation, tools or resources evaluation, etc);
	analyis and adaption of relevant research-literature with respect
	to the topic. It is possible to offer a praxis oriented tutoriums;
	the workload is then the sum of project and tutorium; such tutori-
	ums can be used to mediate, to train and to deepen programming
	techniques, annotation-schemes and -methods, special statistical
	procedures, evaluation methods or usage of complex systems.
Submodules	none
Literature	depending on the topic
Requirements	The contents of the Module "Seminar Computer Linguistic Re-
	sources"
Exam	written summary
Recommended Term	MSc 3
Turn	irregular
Duration	1 Semester
Use	
	• MSc AInf Variant: Data Analytics (fundamental track) –
	Elective Modules – Application – Natural Language Process-
	ing
Language	English

Module: Project Computer Linguistic Processes

Responsible	Prof. Dr. Ulrich Heid
Responsible Instructors	Prof. Dr. Ulrich Heid and members of the study group
Type	4 HPW project
Credit Points	6 CPs
Workload	presence: 42 hours; self-study: 108 hours
Learning goals/ Compe-	The students are able to analyse and evaluate Computer Linguis-
tencies	tic Processes or Language Technology tools and applications. The students should be able to independently solve, analyse and im-
	plement or respectively adapt or optimize questions from the area
	of Compuer Linguistic Processes or Language Technology respec- tively. They are able to relate their solution to state-of-the-art
	research. This seminar gives the foundations for the master thesis.
Submodules Literature	Specialization and integrated, theoretical, methodological and practical research-based treatment of topics in Computer Linguistic and Language technology with focus on Processes. The students independently solve the task with Computer Linguistic resources (if necessary in the case of joint projects like participate shared tasks, resource creation, tools or resources evaluation, etc); analyis and adaption of relevant research-literature with respect to the topic. It is possible to offer a praxis oriented tutoriums; the workload is then the sum of project and tutorium; such tutoriums can be used to mediate, to train and to deepen programming techniques, annotation-schemes and -methods, special statistical procedures, evaluation methods or usage of complex systems. none depending on the topic
Requirements	The contents of the Module "Seminar Computer Linguistic Processes"
Exam	written summary
Recommended Term	MSc 3
Turn	irregular
Duration	1 Semester
Use	MSc AInf Variant: Data Analytics (fundamental track) – Elective Modules – Application – Natural Language Processing
Language	English

Module: Lab Course Natural Language Processing

Responsible	Dr. Folker Caroli
Lecturer	Dr. Folker Caroli and members of the study group
Type	2 HPW lab course
Credit Points	4 CPs
Workload	presence: 21 hours; self-study: 79 hours
Learning goals/ Competencies	Testing of and independently work with tools, methods and resources in Language Processing; evaluation of implementation approaches; implementation- and test/evaluation-practice. Knowledge of relevant tools and programming languages; skills for their productive usage: ability to estimate the realization of practical tasks in Language Processing.
Content	In the Praktikum students deal with concrete research- and development topics in the sense of case studies with a high percentage of practical content. Obtain knowledge and skills with tools, methods and Language Technology resources. For example: • (a) corpus-linguistic tools: tools for the whole corpus-
	 (a) corpus-linguistic tools. tools for the whole corpuslinguistic process, starting with data acquisition and conversion (crawler, scripts) over annotation (tokenizing, tagging, parsing) till data extraction (search tools and strategies). Implementation of simple questions of language and information research in corpus-linguistic analysis; interpretation of the results. (b) script languages for working wih text data: Introduction to one script language (Perl, Python), concepts of programming with script languages, working with big data. Implementation concepts, test and evaluation of scripts. (c) data bases and Natural Language Processing software projects: principles and praxis of relational data bases; SQL, definition and creation of data bases, linguistic data retrieval; adapt data bases in bigger software projects od Language Processing:; architeture and implementation strategies for the work with big text data. Aspects of other data bases approaches (e.g. XML-data bases, object oriented data bases) (d) statistic methods for Language Technology: need and task formulation for descriptive statistic methods for working with language data (e.g. corpus-linguistic work); fundamental statistical methods for distribution calculation, assoziations, evaluation matching, etc. Implementation concepts of statistical programming languages like R or Python.
Submodules	none
Literature	depending on the topic
Requirements	Contents of the lecture "Introduction Natural Language Processing".
Exam	colloquium and written summary
Recommended Term	MSc 2-3

Turn	every year
Duration	1 Semester
Use	• MSc AInf Variant: Data Analytics (fundamental track) – Elective Modules – Application – Natural Language Processing
Language	English

Psychology

Module: Cognitive Psychology

Responsible	Professor Dr. habil. Christina Bermeitinger
Responsible Instructors	Professor Dr. habil. Christina Bermeitinger and members of the
	study group
Type	2 HPW lecture, 2 HPW tutorial
Credit Points	6 CPs
Workload	presence: 42 hours; self-study: 108 hours
Learning goals/ Compe-	After completion of this module, students will have developed a
tencies	deep understanding of concepts and methods of cognitive psychol-
	ogy.
Content	Introduction to the concepts and methods of cognitive psychology.
Submodules	SM 1: Cognitive Psychology, Lecture Type: 2 HPW Lecture (3
	CPs) Lecturer: Professor Dr. habil. Christina Bermeitinger SM
	2: Cognitive Psychology, Tutorium Type: 2 HPW Tutorium (3
	CPs) Lecturer: Professor Dr. habil. Christina Bermeitinger and
	members of the study group
Literature	
	• Eysenck, Michael W., Keane, Mark T. (2020): Cognitive
	Psychology: A Student's Handbook, Taylor and Francis Ltd.
	1 by choice 81. The condition of the configuration and trained from
Requirements	None
Exam	written exam (120 min) or an oral exam (30 min)
Recommended Term	MSc 1-3
Turn	every winter term
Duration	1 Semester
Use	
	MSc AInf Variant: Data Analytics (fundamental track) –
	Elective Modules – Application – Psychology
Language	English

Module: Brain and Neural Science

Responsible	Professor Dr. rer. nat. Kristian Folta-Schoofs
Responsible Instructors	Professor Dr. rer. nat. Kristian Folta-Schoofs and members of the
	study group
Type	2 HPW lecture, 2 HPW tutorial
Credit Points	6 CPs
Workload	presence: 42 hours; self-study: 108 hours
Learning goals/ Compe-	After completion of this module, students will have developed a
tencies	deep understanding of the functional organization and major prop-
	erties of neural circuits of the brain that give rise to perceptions,
	thoughts, and planned actions. They will be able to outline basic
	neural principles essential for understanding human cognition and
	behavior. Furthermore, they will be able to discuss, how sensa-
	tions are assembled into inner representations and formulated into
	plans for immediate behavior or concepts for future actions.

Content	The human brain is a complex network of nerve cells and glial
Content	The human brain is a complex network of nerve cells and glial cells interconnected in systems of the central nervous system which construct our perceptions of the external world, fix our attention, and control our actions. Advances in neural science have been matched by advances in our understanding of the biology of higher brain functions which allow humans to perceive, act, think, speak, learn, and remember. Neural science represents a merger of molecular biology, neurophysiology, anatomy, embryology, cell biology, immunology, and biological psychology. Powerful research tools permit to investigate and describe the functional organization and system properties of neural circuits of the brain (in controlled experiments), and to link the molecular dynamics of individual nerve cells to representations of perceptual and motor acts. This module focuses on major neural structures, principles and mechanisms,
	and their relation to particular modes of cognition and behavior. We attempt to address the following questions:
	• How does the central nervous system and its organization develop?
	• What are the molecular mechanisms underlying the developmental processes responsible for assembling functional neural circuits?
	How do nerve cells and glial cells communicate and interact with one another? How do nerve cells and glial cells communicate and interact with one another?
	 How do properties of specific nerve cells and different patterns of interconnections give rise to higher brain functions and particular modes of thinking and behavior? To what extent are mental processes hard-wired into the neural architecture of the brain?
	 How is communication between neurons influenced by the environment, including the actions of other people? How does the brain and its functional organization produce the remarkable individuality of human cognition and action? What contributes to the pathogenesis of diseases that affect the structures and functions of the nervous system?
Submodules	SM 1: Brain and Neural Science, Lecture Type: 2 HPW Lecture (3 CPs) Lecturer: Professor Dr. rer. nat. Kristian Folta-Schoofs SM 2: Brain and Neural Science, Tutorium Type: 2 HPW Tutorium (3 CPs) Lecturer: Professor Dr. rer. nat. Kristian Folta-Schoofs and members of the study group
Literature	• Eric R. Kandel, John D. Koester, Sarah H. Mack, Steven A Siegelbaum: Principles of Neural Science. Sixth Edition. McGraw Hill, 2021.
Requirements	None
Exam	written exam (120 min) or an oral exam (30 min)
Recommended Term	MSc 1-3

Turn	every summer term
Duration	1 Semester
Use	• MSc AInf Variant: Data Analytics (fundamental track) – Elective Modules – Application – Psychology
Language	English

Environmental Sciences

Module: Geographic Information Systems

Responsible	Prof. Dr. Martin Sauerwein
Type	2 HPW lecture, 4 HPW tutorial
Credit Points	6 CPs
Workload	presence: 63 hours; self-study: 87 hours
Learning goals/ Competencies	The students should be able to visualize geographic data and creating professional maps using GIS (SM 1). The second aim is that students gain basic knowledge of geoprocessing tools for the analysis of spatial geographic data in GI-Systems (SM 2).
Content	
	 SM 1: The course introduces the fundamental concepts of Geographic Systems (GIS) and includes computer exercises with the software ArcGIS. In the course students acquire practical skills of major work steps as creating and editing spatial data, integrating data from different data sources, georeferencing, spatial queries, and creating and exporting maps. SM 2: The second course focuses on spatial analysis and geoprocessing tools. In different GIS-projects real data sets are used for a spatial analysis with the software ArcGIS. The integration of software knowledge and ecological knowledge will be practiced.
Submodules	SM 1: Geographic Information Systems I Type: 2 HPW Lecture (2 CPs) 1 HPW Tutorium (1 CP) Lecturer: Prof. Dr. Martin Sauerwein SM 2: Geographic Information Systems II Type: 2 HPW Tutorium (3 CPs) Lecturer: Prof. Dr. Martin Sauerwein
Literature	will be announced in the lecture
Requirements	none
Exam	module exam: submission of a GIS-project and oral presentation of the GIS-project (20 min)
Recommended Term	MSc 1-3
Turn	every year
Duration	2 Semester
Use	MSc AInf Variant: Data Analytics (fundamental track) – Elective Modules – Application – Environmental Sciences
Language	English

Soft Skills

Module: English 1

Responsible	external lecturer
Type	2 HPW lecture
Credit Points	3 CPs
Workload	presence: 21 hours; self-study: 54 hours
Learning goals/ Compe-	Students should be able to communicate and negotiate in correct
tencies	English especially in the area of Data Analytics. They should know
	the specialist vocabular for written and oral communication.
Content	Writing and speaking English. Learn English grammar.
Submodules	SM 1: English 1 I Type: 2 HPW Lecture (3 CPs) Lecturer: ex-
	ternal lecturer SM 2: English 1 II Type: 2 HPW Lecture (3 CPs)
	Lecturer: external lecturer
Literature	Refer to instructor
Requirements	none
Exam	Refer to instructor
Recommended Term	MSc 1-3
Turn	every winter term
Duration	1 Semester
Use	
	• MSc AInf Variant: Data Analytics (fundamental track) –
	Elective Modules – Application – Soft Skills
Language	English

Module: English 2

Responsible	external lecturer
Type	2 HPW lecture
Credit Points	3 CPs
Workload	presence: 21 hours; self-study: 54 hours
Learning goals/ Compe-	Students should be able to run specialist negotiations in correct
tencies	English, to speak and discuss fluently. They earn a great vocabulary
	to negotiate in the area of Data Analytics. They know the formal
	business ommunication and what has to be respected. Students
	should be able to present their results in English.
Content	Writing and speaking English. Learn English grammar.
Submodules	SM 1: English 2 I Type: 2 HPW Lecture (3 CPs) Lecturer: ex-
	ternal lecturer SM 2: English 2 II Type: 2 HPW Lecture (3 CPs)
Lecturer	external lecturer
Literature	Refer to instructor
Requirements	Refer to instructor
Exam	Refer to instructor
Recommended Term	MSc 1-3
Turn	every summer term
Duration	1 Semester
Use	
	• MSc AInf Variant: Data Analytics (fundamental track) –
	Elective Modules – Application – Soft Skills
	••
Language	English

Module: German 1

Responsible	external lecturer
Type	2 HPW lecture
Credit Points	3 CPs
Workload	presence: 21 hours; self-study: 54 hours
Learning goals/ Compe-	This course should encourage students to speak German.
tencies	
Content	This course focuses on different chapters of the German grammar,
	vocabulary and conversation.
Submodules	SM 1: German 1 I Type: 2 HPW Lecture (3 CPs) Lecturer:
	external lecturer SM 2: German 1 II Type: 2 HPW Lecture (3
	CPs) Lecturer: external lecturer
Literature	Given in Class
Requirements	none
Exam	Exam
Recommended Term	MSc 1-3
Turn	every winter term
Duration	1 Semester
Use	
	• MSc AInf Variant: Data Analytics (fundamental track) –
	Elective Modules – Application – Soft Skills
	Tr
Language	English/German

Module: German 2

Responsible	external lecturer
Type	2 HPW lecture
Credit Points	3 CPs
Workload	presence: 21 hours; self-study: 54 hours
Learning goals/ Compe-	This course should encourage students to speak German.
tencies	
Content	This course focuses on different chapters of the German grammar,
	vocabulary and conversation.
Submodules	SM 1: German 2 I Type: 2 HPW Lecture (3 CPs) Lecturer:
	external lecturer SM 2: German 2 II Type: 2 HPW Lecture (3
	CPs) Lecturer: external lecturer
Literature	Refer to instructor
Requirements	German 1 or equivalent German knowledge
Exam	Refer to instructor
Recommended Term	MSc 1-3
Turn	every summer term
Duration	1 Semester
Use	
	MSc AInf Variant: Data Analytics (fundamental track) –
	Elective Modules – Application – Soft Skills
	**
Language	English/German

Module: International Start-up School

	SA
Responsible Instructors Diana Püplichhuysen, M.A., MB	SA
Lecturer Diana Püplichhuysen, M.A., MB	SA
Type 2 HPW lecture	
Credit Points 3 CPs	
Workload presence: 21 hours; self-study: 54	4 hours
tencies to set up a business in German their business idea into practice while considering the specifics of the legal and tax regulations for sthermore students are made fami of start-up financing an the supp	al students the core competencies by. Students are enabled to put by means of business modelling of the technology sector as well as self employment in Germany. Furliar with the different possibilities port structures offered by regional
networks.	
ship 3. Fundamentals of the Legal ulations for self-employed 4. Start-up Financing 5. Start-up-related networks i	Sector and Digital Entrepreneur- System and Corporate Tax Reg- in the region
(3 CPs) Lecturer: external lect	School I Type: 2 HPW Lecture surer SM 2: International Start-cture (3 CPs) Lecturer: external
Literature none	
Requirements none	
Exam Refer to instructor	
Recommended Term MSc 1-3	
Turn every winter term	
Duration 1 Semester	
Use • MSc AInf Variant: Data Elective Modules – Applica	Analytics (fundamental track) – ation – Soft Skills

Master Thesis

Module: Master Thesis

Responsible	Prof. Dr. Lars Schmidt-Thieme
Lecturer	According to agreement
Type	Master Thesis
Credit Points	30 CPs
Workload	presence: 0 hours; self-study: 750 hours
Learning goals/ Compe-	Student will acquire in-depth knowledge of the area data analytics.
tencies	Through their master thesis projects students will gain necessary
	research skills and expertise on a specific issue of their own choosing
	in the area of data analytics.
Content	Students demonstrate adequate knowledge and understanding that
	provides a basis for developing original ideas within an academic
	context. Integrate knowledge and processes complex information
	that link to the chosen research topic. They prove their ability in
	reporting, source-finding, critical thinking and analysis, problem
	formulation and solving, argumentation, and reasoning.
Submodules	SM 1: Master Thesis, Lecture Type: written thesis (27 CPs) Lec-
	turer: Professors in the course of study SM 2: Master Thesis,
	Tutorium Type: colloquium (3 CPs) Lecturer: Professors in the
	course of study
Literature	depends on the topic
Requirements	passed all necessary modules
Exam	written thesis
Recommended Term	MSc 4
Turn	every semester
Duration	1 Semester
Use	
	MSc AInf Variant: Data Analytics (fundamental track) –
	• MSc AInf Variant: Data Analytics (fundamental track) –
	Master Thesis
Language	English