## Handbook of Modules

for the

### Master Study Programme Data Science

(Terms of study 2019/2020)

#### updated July 2023

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Module MD E1-6: Robust Methods	
Module MD E1-7: Optimization	
Module MD E1-8: Numerical Solution of Differential Equations	
Module MD E1-9: Generalized Linear Models	
Module MD E1-10: Introduction to Computational Fluid Dynamics	
Module MD E1-11: Introduction to Computational Intelligence	
Module MD E1-12: Architecture and Implementation of Database Systems	
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Module MD E1-14: Real-Time Systems and Applications	
Module MD E1-15: Model Selection and Model Averaging	
Module MD E1-16: Asymptotic Theory	
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Module MD E1-18: Neural networks for solving ordinary differential equations	
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Module MD E1-20: Artificial Intelligence	
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Module MD E2-1: Bioinformatics	
Module MD E2-2: Toxicology	
Module MD E2-3: Econometrics	
Module MD E2-4: Econometrics of treatment effects and policy evaluation	
Module MD E2-5: Natürlichsprachige Systeme	
Module MD E2-6: Control Theory and Applications	
Module MD E2-7: Reliability and Material Fatigue	
Module MD E2-8: Quality Control	
Module MD E2-9: Financial Econometrics	
Module MD E2-10: Machine Learning for Economic Data	
Module MD E2-11: Bayesian Econometrics	
Module MD E2-12: Statistics in Toxicology (Modelling)	
Module MD E2-13: Statistics in Toxicology (Testing)	
Module MD E2-14: Statistics in Genetics (Bioinformatics)	
Module MD E2-15: Introduction to Linguistic Data Science with R	
Module MD E2-16: Introduction to Computational Linguistics with Python	
Module MD E2-17: Statistics meets Linguistics	
Module MD E2-18: Research Methods in English Linguistics	
Module MD E2-19: Learning in Robotics	
Module MD E2-20: Data-Based Dynamic Modeling	
Module MD E2-21: Industrial Data Science I	
Module MD E2-22: Game Theory	
Module MD E2-23: Introduction to Linguistic Data Science	
Module MD E2-24: Introduction to Computational Linguistics	
Possible requirements in case of conditional admission	
Module MD Req1: Advanced Mathematics	
Module MD Req2: Data Structures and Programming	
Module MD Req3: Information Systems	45
Module MD Req4: Probability	
Module MD Req5: Inference	
Module MD Req6: Linear Models	
Module MD Req7: Introductory Case Studies	50
Module MD Req8: Introduction to Statistical Learning	

#### Overview of modules

#### Compulsory Courses

No.	MODULE NAME	Lectures/courses	ECTS points
MD 1	ADVANCED STATISTICAL LEARNING	Advanced Statistical Learning	9
MD 2	STATISTICAL THEORY	Statistical Theory	9
MD 3	DATA SCIENCE IN PRACTICE	Programming course	3
		Data Science in Context	3
MD 4	PROJECT WORK	Case Studies / External Internship	8
		Seminar	4
MD 5	BIG DATA	Statistical Learning for Big Data	9
MD 6	MASTER THESIS	Master Thesis	27
		Advanced Seminar	3

#### Elective courses

(to choose modules with a total of 45 ECTS points)

CATALOGUE OF MODULES	No.	Lectures/courses	ECTS points	Language
	MD E1-1	Time Series Analysis	9	English
	MD E1-2	Survival Analysis	9	English
	MD E1-3	Bootstrapping	9	English
	MD E1-4	Stichprobenverfahren	5	German
	MD E1-5	Nonparametric Methods	4,5	English or German
	MD E1-6	Robust Methods	4,5	English or German
	MD E1-7	Optimization	5	English
METHODS (to choose modules with a	MD E1-8	Numerical Solution of Differential Equations	5	English
total of at least 24 ECTS	MD E1-9	Generalized Linear Models	9	English
points)	MD E1-10	Introduction to Computational Fluid Dynamics	5	English
	MD E1-11	Introduction to Computational Intelligence	4	English
	MD E1-12	Architecture and Implementation of Database Systems	8	English
	MD E1-13	Cyber-Physical System Fundamentals	6	English
	MD E1-14	Real-Time Systems and Applications	8	English
	MD E1-15	Model Selection and Model Averaging	9	English
	MD E1-16	Asymptotic Theory	4,5	English

MD E1-17	Machine Learning Paradigms for Complex	8	English
MD E1-18	Neural networks for solving ordinary	5	English
MD E1-10	· · ·	5	English
			English
	-		English
-	, ,		English
	- · ·		English
			English
		-	English
-			English
	Causanty	0	English or
Other modu	ules upon request		German
MD E2-1	Bioinformatics	9	English
MD E2-2	Toxicology	9	English
MD E2-3	Econometrics	9	English
MD E2-4	Econometrics of treatment effects and policy evaluation	4,5	English
MD E2-5	Natürlichsprachige Systeme	7	German
MD E2-6	Control Theory and Applications	7	English
MD E2-7	Reliability and Material Fatigue	9	English or German
MD E2-8	Quality Control	9	English or German
MD E2-9	Financial Econometrics	4.5	English
		-	English
	_		English
		-	English
		-	English
			English
MD E2-15	Introduction to Linguistic Data Science with R	5	English
MD E2-16	Introduction to Computational Linguistics with Python	5	English
MD E2-17	Statistics meets Linguistics	4,5	English
MD E2-18	Research Methods in English Linguistics	4,5	English
MD E2-19	Learning in Robotics	5	English
MD E2-20	Data-Based Dynamic Modeling	3	English
MD E2-21	Industrial Data Science I	5	English
MD E2-22	Game Theory	7,5	English
MD E2-23	Introduction to Linguistic Data Science	10	English
MD E2-24	Introduction to Computational Linguistics	10	English
Other modu	iles upon request		English or German
	MD E1-18         MD E1-19         MD E1-20         MD E1-21         MD E1-23         MD E1-23         MD E1-24         MD E1-25         MD E1-26         Other modu         MD E2-1         MD E2-3         MD E2-3         MD E2-4         MD E2-3         MD E2-3         MD E2-4         MD E2-3         MD E2-4         MD E2-3         MD E2-10         MD E2-10         MD E2-10         MD E2-11         MD E2-10         MD E2-11         MD E2-13         MD E2-14         MD E2-23         MD	DataDataMD E1-18Neural networks for solving ordinary differential equationsMD E1-19Numerical Methods for PDEsMD E1-20Artificial IntelligenceMD E1-21Deep LearningMD E1-22Unit Root and Cointegration AnalysisMD E1-23Panel Data EconometricsMD E1-24Text as DataMD E1-25Statistical Network AnalysisMD E1-26CausalityOther modules upon requestMD E2-1BioinformaticsMD E2-2ToxicologyMD E2-3Econometrics of treatment effects and policy evaluationMD E2-4Econometrics of treatment effects and policy evaluationMD E2-5Natürlichsprachige SystemeMD E2-6Control Theory and ApplicationsMD E2-7Reliability and Material FatigueMD E2-8Quality ControlMD E2-9Financial EconometricsMD E2-11Bayesian EconometricsMD E2-12Statistics in Toxicology (Modelling)MD E2-13Statistics in Genetics (Bioinformatics)MD E2-14Statistics in Genetics (Bioinformatics)MD E2-15Introduction to Computational LinguisticsMD E2-16Introduction to Computational LinguisticsMD E2-17Statistics meets LinguisticsMD E2-18Research Methods in English LinguisticsMD E2-19Learning in RoboticsMD E2-21Industrial Data Science IMD E2-22Game Toxicology (Modelling)MD E2-23Introduction to Linguistic Data Science <td>DataData8MD E1-18Deural networks for solving ordinary differential equations5MD E1-19Numerical Methods for PDEs5MD E1-20Artificial Intelligence3MD E1-21Deep Learning9MD E1-22Unit Root and Cointegration Analysis9MD E1-23Panel Data Econometrics9MD E1-24Text as Data4,5MD E1-25Statistical Network Analysis4,5MD E1-26Causality6Other modules upon request9MD E2-1Bioinformatics9MD E2-2Toxicology9MD E2-3Econometrics of treatment effects and policy evaluation4,5MD E2-4Econtrol Theory and Applications7MD E2-5Natürlichsprachige Systeme7MD E2-6Control Theory and Applications7MD E2-7Reliability and Material Fatigue9MD E2-8Quality Control9MD E2-9Financial Econometrics4,5MD E2-10Machine Learning for Economic Data4,5MD E2-11Bayesian Econometrics4,5MD E2-12Statistics in Toxicology (Modelling)4,5MD E2-13Statistics in Toxicology (Testing)4,5MD E2-14Introduction to Computational Linguistics5MD E2-15Introduction to Computational Linguistics4,5MD E2-16Introduction to Computational Linguistics4,5MD E2-17Statistics metes Linguistic Data Science5&lt;</td>	DataData8MD E1-18Deural networks for solving ordinary differential equations5MD E1-19Numerical Methods for PDEs5MD E1-20Artificial Intelligence3MD E1-21Deep Learning9MD E1-22Unit Root and Cointegration Analysis9MD E1-23Panel Data Econometrics9MD E1-24Text as Data4,5MD E1-25Statistical Network Analysis4,5MD E1-26Causality6Other modules upon request9MD E2-1Bioinformatics9MD E2-2Toxicology9MD E2-3Econometrics of treatment effects and policy evaluation4,5MD E2-4Econtrol Theory and Applications7MD E2-5Natürlichsprachige Systeme7MD E2-6Control Theory and Applications7MD E2-7Reliability and Material Fatigue9MD E2-8Quality Control9MD E2-9Financial Econometrics4,5MD E2-10Machine Learning for Economic Data4,5MD E2-11Bayesian Econometrics4,5MD E2-12Statistics in Toxicology (Modelling)4,5MD E2-13Statistics in Toxicology (Testing)4,5MD E2-14Introduction to Computational Linguistics5MD E2-15Introduction to Computational Linguistics4,5MD E2-16Introduction to Computational Linguistics4,5MD E2-17Statistics metes Linguistic Data Science5<

#### Possible requirements in case of conditional admission

No.	MODULE NAME	Lectures/courses	ECTS points
MD Req1	ADVANCED MATHEMATICS	Advanced Engineering Mathematics	7
MD Req2	DATA STRUCTURES AND PROGRAMMING	Reading Course Data Structures and Programming	10
MD Req3	INFORMATION SYSTEMS	Reading Course Information Systems	5
MD Req4	PROBABILITY	Reading Course Probability	5
MD Req5	INFERENCE	Reading Course Inference	5
MD Req6	LINEAR MODELS	Reading Course Linear Models	5
MD Req7	INTRODUCTORY CASE STUDIES	Introductory Case Studies	5
MD Req8	INTRODUCTION TO STATISTICAL LEARNING	Introduction to Statistical Learning	10

#### Notes

- Elective courses can be in English or German (see terms of study, § 6 (6)).
- For courses with a written exam according to this handbook, there have to be two examination dates during the semester, generally one of them towards the end of the free period. Students not having passed or participated in the first date may participate in the second. If the examination is not passed at the second date, there is no right to a further attempt during the same semester, even if the student has not written the exam at the first regular date. For courses that are offered every semester, and not only once per year, it is possible to have only one exam after each lecture period, resulting in a total of two exams per year.
- Course type abbreviations:
  - o L: Lecture
  - $\circ \quad \text{T: Tutorial}$
  - o S: Seminar
  - P: Practical course

#### Preface on dissemination of skills

Understanding of the methodology of mathematical statistics cannot be acquired by simple memorising, but solely by exercises and application to practical questions and problems. This is usually done in tutorials accompanying all lectures. Therefore, they are a central part of the education.

Apart from understanding of the lectures' contents, the accompanying tutorials impart strategies of learning: the students recognise the application of methods to practical questions as the easiest way of understanding. Writing out answers and solutions to exercises improves the ability to formulate propositions of mathematical statistics and the general capability of expression.

Application to real data is only possible by means of software. Therefore, after basic training in earlier phases of study, computational application of studied methods is required in the courses. Self-discipline and time management are trained by periodic and strict deadlines for finishing exercises. Motivation is increased by feedback on corrected exercises, since the students can judge their understanding of the subjects worked on. Therefore, it is an important part of the education to value and reward the students' efforts in doing exercises.

Various examination modes are provided in the regulations. In particular, credit points may be obtained by the exercises mentioned above and by oral presentations. Communication skills and the ability to describe issues of mathematical statistics and data science are trained by oral exams. Written exams require the methods' practical application.

Perhaps the most important soft skills in statistics and data analysis are communication and teamworking skills. Expedient data analysis and evaluation can only be achieved if all aspects of the data and problems are openly discussed between all the participants. The department of Statistics encourages these abilities by instructions to teamwork in various ways and by special tutorials held by fellow students.

A semester abroad also serves to develop such general, interdisciplinary skills. Students are encouraged to take some courses at a foreign partner university. In particular, the 3rd semester is suitable in this regard. Such international exchanges are for example supported by the ERASMUS programme (https://www.statistik.tu-dortmund.de/erasmus\_en.html).

#### Recommended course of study

#### Starting in winter semester

1st semester	2nd semester	3rd semester	4th semester
Module MD 5:	Module MD 4:		Module MD 6:
Big Data	Project Work		Master Thesis
Statistical Learning for Big Data (4+2); 9	Case Studies (4P) or	Seminar (2S); 4 ECTS;	Prerequisites: Modules MD 1 and
ECTS;	External Internship; 8 ECTS;	Graded partial work	MD 4
Graded module exam	Graded partial work		
			30 ECTS;
Module MD 2:	Module MD 1:		Course achievement: Advanced Seminar
Statistical Theory	Advanced Statistical Learning		("Oberseminar"); Graded module exam: Master Thesis
Statistical Theory (4+2); 9 ECTS;	Advanced Statistical Learning (4+2); 9	Graded module exams or accumulated	
Graded module exam	ECTS;	graded exams	
	Graded module exam		
	Elective courses:	(at least 24 ECTS in total)	
	Methods		
	inclidus	(In the entire elective area – Methods and	
Module MD 3:	Elective modules from catalogue	Applications – modules with a total of 45	
Data Science in Practice		ECTS are to be chosen.)	
		Let's the to be chosen.y	
Programming course (2 to 4 P); 3 ECTS;	Elective courses:		
/	Applications	Graded module exams or accumulated	
Data Science in Context (2); 3 ECTS;		graded exams	
	Elective modules from catalogue		
Accumulated graded exams		(at least 16 ECTS in total)	
Total: 30 ECTS	Total: 30 ECTS	Total: 30 ECTS	Total: 30 ECTS

blue: courses at the Faculty of Statistics green: courses at the Faculty of Computer Science brown: (joint) courses at this or other faculties Denoted hours: S: Seminar P: Practical course

else: Lecture + Tutorial or Lecture only

#### Starting in summer semester

1st semester	2nd <u>semester</u>	3rd <u>semester</u>	4th <u>semester</u>
Module MD 1: Advanced Statistical Learning	Module MD 4: Project Work	Case Studies (4P) or	Module MD 6: Master Thesis
Advanced Statistical Learning (4+2); 9 ECTS:	Seminar (2S); 4 ECTS;	External Internship; 8 ECTS; Accumulated graded exams	Prerequisites: Modules MD 1 and MD 4
Graded module exam			
Elective courses: Methods	Module MD 5: Big Data	Graded module exams or accumulated	30 ECTS; Course achievement: Advanced Seminar ("Oberseminar");
Elective modules from catalogue	Statistical Learning for Big Data (4+2); 9 ECTS; Graded module exam	graded exams (at least 24 ECTS in total)	Graded module exam: Master Thesis
	Module MD 2: Statistical Theory		
	Statistical Theory (4+2); 9 ECTS; Graded module exam	(In the entire elective area – Methods and Applications – modules with a total of 45	
		ECTS are to be chosen.)	
Module MD 3: Data Science in Practice	Elective courses: Applications	Elective modules from catalogue	
Programming course (2 to 4 P); 3 ECTS;	Data Science in Context (2); 3 ECTS;	Graded module exams or accumulated graded exams	
Accumulated graded exams		(at least 16 ECTS in total)	
Total: 30 ECTS	Total: 30 ECTS	Total: 30 ECTS	Total: 30 ECTS

blue: courses at the Faculty of Statistics green: courses at the Faculty of Computer Science brown: (joint) courses at this or other faculties

Denoted hours: S: Seminar

P: Practical course else: Lecture + Tutorial or Lecture only

#### Description of modules

#### Compulsory Courses

Module MD 1: Advanced Statistical Learning						
Rota	Duration	Semester	ECTS points	Workload		
Summer semester, annual	1 semester	1st	9	270		

1	Module	Module structure						
	No.	Lecture/Course	Туре	ECTS	Hours			
	1	Advanced Statistical Learning	L+T	9	4 + 2			
2	Language	9						
	English							
3	Content							
	-	cts of the statistical learning beginner		-	-			
	•	BD 10) are enhanced and expanded.						
		neural networks (deep learning), adv	-					
	-	ed additive (mixed) models, enhance	• •	-				
		are introduced. On the other hand, t						
		are shown, in particular, with regard ntal differences between supervised	-					
4		ences / Qualification Goals	and non-supervi	seu learning	are evolveu.			
4	•	ents understand advanced models an	d analysing metl	hods and are	aware of their			
		ns. They are able to adapt methods to						
		ate methods for real data and apply t	•		•			
		nd the underlying mathematical theo	•					
5	Examina	tion	-					
		nodule exam.						
6		y of examination						
	Written	or oral exam as announced by the lect	turer.					
7	Modula	requirements (prerequisites)						
<b>′</b>	- none -	requirements (prerequisites)						
8		n to curriculum						
		ory module for M. Sc. study program	ne Data Science					
9	Respons		Department					
	•	K. Ickstadt, Prof. Dr. J.	Statistics					
		ührer, Prof. Dr. A. Groll						
L								

Module MD 2: Statistical Theory						
Rota	Duration	Semester	ECTS points	Workload		
Winter semester, annual	1 semester	1st	9	270		

1	1 Module structure					
	No.	Lecture/Course	Туре	ECTS	Hours	
	1	Probability Theory	L+T	4,5	2 + 1	
	2	Decision Theory	L+T	4,5	2 + 1	
2	Langua	ge	·			
	English					
3	Conten	t				
		itistical Theory lecture (6 + 3) is divid	• •		•	
		n Theory are compulsory, while the	third part (Asympto	otic Theory) n	nay be chosen for	
		r module.				
		bability Theory part deals with basi	•		eory to enable	
		ts for the comprehension of more so	•	cal methods.		
	The De	cision Theory part deals with basics	of decision theory.			
4	-	tences / Qualification Goals				
		idents handle the formal language of				
		ility theory, decision theory and ma		s. They are we	ell prepared to	
	study, o	correctly apply and enhance statistic	cal methods.			
5	Examir	ation				
	Graded	l module exam.				
_						
6		ity of examination				
		n exam on the compulsory parts.				
		turer may demand course achievem	ients as a prerequis	ite to the exa	m, if announced	
_		beginning of the course.				
7		e requirements (prerequisites)				
_	- none					
8		ion to curriculum	<b>-</b>			
		Isory module for M. Sc. study progra				
9	-	nsibility	Department			
	Prof. D	r. K. Ickstadt, Prof. Dr. C. Jentsch	Statistics			

Mc	Module MD 3: Data Science in Practice								
Ro	ta		Duration	Semester		ECTS points		Workload	
anı	nual		1 to 2	1st to 2nd	l	6		180	
			semesters						
1	Module structure								
	No. Lecture/Course			Туре		ECTS	Hours		
	1	_	nming course		Р		3	2 to 4	
	2		ience in Context		L		3	2	
2	Language								
	English o	r Germar	ו						
3	Content								
			ts participate in a s						
	•		dvanced R, which t						
			includes introduct elated to data (data						
	0		ngst others:	i privacy ari	u mtegnty)	, and to	o the prac	lical aspects of	
			nagement (RDS) as	nart of the	research r	nncess	ensures a	an efficient	
			ch projects, high q	•	•				
			everal stakeholder						
	•		ata life cycle contai				-		
	-		ving, and sharing of	• .				-	
	RDS and	present h	nands on examples	from every	day resear	ch. Furt	thermore,	the usefulness	
	of databa	ises in th	e context of RDS fo	or placemer	nt and inqui	iry of re	esearch da	ata will be shown,	
	with a sp	ecial focu	us on format divers	ity, curatio	n and acces	ssibility			
4	•	• •	ualification Goals						
			specialized knowle						
	• • •		Is from statistics ar		•	data u	sing their	own program	
			unknown computa		•				
	The stude	ents knov	w the important as	pects of the	eir protessio	on.			
5	Examinat								
			graded exams.						
6	Formality								
	Exams as	annound	ed by the lecturer	S.					
7	Module r	equirem	ents (prerequisite	s)					
	- none -								
8	Allocatio	n to curr	iculum						
	-	-	le for M. Sc. study	programm	e Data Scie	nce			
9	Responsi	-			Departmer	nt			
	Chairmar	n of board	d of examiners		Statistics				

	<b>ta</b> ch semeste	or.	<b>Duration</b> 2 semesters	Semester 2nd to 3rd		<b>ECTS p</b> 12	points	<b>W</b>	orkload	
	offered	:,	2 semesters	2110 10 510		12		50	50	
			·							
1	Module s						5.070			
	No.		/Course			Type ECTS		Hours		
	1a	Case St			Р		8		4	
	1b		l Internship		6		8		2	
-	2	Semina	r		S		4		2	
2	Language									
		- ·	t least once per y							
			German, by arra German, by arran							
3	Content	glish of C	Serifiali, Dy allali	igement						
5		the stu	dents choose bet	tween internal a	nd ovtor	nal ovo	rcisos in	nron	aration for	
	professio							prep		
			 es" course, the s	tudents work or	one or	two ext	ensive n	roied	cts dealing	
			lications. They w				•	-	0	
			se, adopt knowle	•		-			-	
		•	•						•	
	-	-	•	•	for the given problem. They write reports and make presentations on the methodol					
	their conducted extensive data analysis. They practise interactions with other field						tields of			
	science and application. During the external internship, the students work at a company or institution related t							ner t	fields of	
		• •								
	During th	e extern	al internship, the	e students work	at a com	ipany or	r institut	ion r	elated to	
	During th their field	e extern I of study	al internship, the y. They participat	e students work te in professiona	at a com	ipany or	r institut	ion r	elated to	
	During th their field analyses.	e extern l of study They rep	al internship, the	e students work te in professiona t and results.	at a com Il workin	ipany or g group	r institut os condu	ion r cting	elated to greal data	
	During th their field analyses. In the ser	e extern l of study They rep ninar, ev	al internship, the y. They participat port their project	e students work te in professiona t and results. Is a scientific pa	at a com Il workin per, criti	ipany or g group cally ref	r institut os condu lecting i	ion r cting ts co	elated to greal data ntents. The	
4	During th their field analyses. In the ser main resu	e extern I of study They rep ninar, ev ults of th	al internship, the y. They participat port their project very student reac	e students work te in professiona t and results. Is a scientific pa ented in a talk, d	at a com Il workin per, criti	ipany or g group cally ref	r institut os condu lecting i	ion r cting ts co	elated to greal data ntents. The	
4	During th their field analyses. In the ser main resu <b>Compete</b>	e extern I of study They rep ninar, ev ults of th <b>nces / Q</b>	al internship, the y. They participat port their project very student reac is work are prese	e students work te in professiona t and results. Is a scientific papented in a talk, d Is	at a com I workin Der, criti iscussed	ipany or g group cally ref and rep	r institut os condu flecting if ported ir	ion r cting ts con n writ	elated to greal data ntents. The tten form.	
4	During th their field analyses. In the ser main resu <b>Compete</b> The stude	e extern I of study They rep ninar, ev <u>ults of th</u> nces / Q ents wor	al internship, the y. They participat port their project very student reac is work are prese ualification Goa	e students work te in professiona t and results. ds a scientific pap ented in a talk, d <b>Is</b> on a given probl	at a com Il workin Der, criti iscussed em, coll	ipany or g group cally ref and rep aboratir	r institut os condu ilecting i ported in ng in tea	ion r cting ts con n writ ms. 1	elated to greal data ntents. The tten form. They adopt	
4	During th their field analyses. In the ser main resu <b>Compete</b> The stude previousl	e extern I of study They rep ninar, ev ults of th nces / Q ents wor y unknow	al internship, the y. They participat port their project very student reac is work are prese ualification Goal k independently	e students work te in professiona t and results. Is a scientific papented in a talk, d Is on a given probl adapt them to r	at a com I workin ber, criti iscussed em, coll eal data	ipany or g group cally ref and rep aboratir tasks. T	r institut os condu ilecting if ported ir ng in tea They inte	ion re cting ts con <u>writ</u> ms. 1 eract	related to g real data ntents. The tten form. They adopt with co-	
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Mc	Module MD 4: Project Work				
8	Allocation to curriculum				
	Compulsory module for M. Sc. study programmer	ne Data Science			
9	Responsibility	Department			
	Chairman of board of examiners	Statistics			

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Module MD 5: Big Data									
Rota	Duration	Semester	ECTS points	Workload					
Winter semester, annual	Winter semester, 1 semester 2nd 9 270								

1	Module	structure							
	No.	Lecture/Course	Туре	ECTS	Hours				
	1	Statistical Learning for Big Data	L+T	9	4 + 2				
2	Language								
	English								
3	Content								
		ourse, methods for handling and a							
		large numbers of both the observ							
		ires generalized for use in a big da	· ·	-					
		ive methods (including cluster and ion and prediction for various clas		•					
		zed linear model, discriminant an							
	-	nachines. Furthermore, methods f	•		•••				
		computing techniques for the afo	-						
		ies of the methods and algorithms		•					
	-	sistency, complexity, etc.		, ,					
4	Compet	ences / Qualification Goals							
	The stud	dents know techniques for dealing	with large data sets	s. They handle	these data and				
		opropriate methods for various tag							
		ues to adapt methods to large dat	•	• •					
		ning for data streams and choose		thods. They k	now how to				
		ze the methods on several cores a	•	-   (I	ha ala sala sala				
	The students apply methods to real data by means of statistical software. They understand								
5	Examina	erlying mathematical theory.							
5		module exam.							
6		ty of examination							
7		announced by the lecturer. requirements (prerequisites)							
1	– none -								
8		on to curriculum							
		sory module for M. Sc. study prog	ramme Data Science	2					
9	Respons	· · · · ·	Department	-					
	•	. E. Müller	Computer Scie	ence					

KO	ta	Duration Semester		Semester		ECTS	points	W	orkload
ea	ch semest	er	1 semester 4th			30	•	90	900
	T								
L		structure							
	No.		/Course		Туре		ECTS		Hours
	1	Master		· //)			27		
	2		ed Seminar ("Ob	erseminar")	S		3		
2	Languag								
3	-	or Germai	1						
)	Content	tor These				امطحيا	coloratif.		thadata -
			s proves the stud						
	•		cience and relat		•			tude	ents may
			several proposed	-				- المار	
			o be written at -						
		•	pecial seminar, the	•					
			outation. This ma		-	inal ta	ік, e.g., a	i a co	ompany or
			nce, in the prese						
1		-	ualification Goa			hice+ ·	onduct -	: ام ر ب	oc and ada
			ependently adop				Lonauct S	scual	es, and adal
		and reflect methods. They communicate the project, methods and results in a well-							
	structured manner and discuss them critically. They compactly present and discuss their							in a	well-
		ed manne	er and discuss th	em critically. T	ney compa	actly pr	esent and	in a d dis	well- cuss their
	own wor	ed manne k. They o	er and discuss th design their proj	em critically. T ect independe	ney compa ntly, toget	actly pr her wit	esent and h the sup	in a d dis	well- cuss their
	own wor perhaps	ed manne k. They o an exterr	er and discuss th	em critically. T ect independe	ney compa ntly, toget	actly pr her wit	esent and h the sup	in a d dis	well- cuss their
5	own wor perhaps <b>Examina</b>	ed manne k. They o an exterr tion	er and discuss th design their proj nal partner, and	em critically. T ect independe finish it within	ney compa ntly, toget	actly pr her wit	esent and h the sup	in a d dis	well- cuss their
	own wor perhaps <b>Examina</b> Graded r	ed manne k. They o an exterr tion nodule e	er and discuss the design their proj nal partner, and xam: Master The	em critically. T ect independe finish it within	ney compa ntly, toget	actly pr her wit	esent and h the sup	in a d dis	well- cuss their
	own wor perhaps <b>Examina</b> Graded r <b>Formalit</b>	ed manne k. They o an exterr tion module e y of exan	er and discuss the design their proj nal partner, and xam: Master The nination	em critically. T ect independe finish it within	ney compa ntly, toget	actly pr her wit	esent and h the sup	in a d dis	well- cuss their
	own wor perhaps <b>Examina</b> Graded r <b>Formalit</b> Successf	ed manne k. They o an exterr tion module e y of exan ul pass of	er and discuss the design their proj nal partner, and xam: Master The nination f thesis.	em critically. T ect independer finish it within esis.	ney compa ntly, toget a given pe	actly pr her wit riod of	esent and h the sup time.	in a d dis bervi	well- cuss their sor and
	own wor perhaps <b>Examina</b> Graded r <b>Formalit</b> Successf Course a	ed manne k. They o an exterr tion module e y of exan ul pass of chieveme	er and discuss the design their proj nal partner, and xam: Master The nination f thesis. ent prior to subn	em critically. T ect independer finish it within esis.	ney compa ntly, toget a given pe	actly pr her wit riod of	esent and h the sup time.	in a d dis bervi	well- cuss their sor and
6	own wor perhaps <b>Examina</b> Graded r <b>Formalit</b> Successf Course a Seminar	ed manne k. They o an exterr tion module e y of exan ul pass of chieveme ("Oberse	er and discuss the design their proj- nal partner, and xam: Master The nination f thesis. ent prior to subn minar").	em critically. T ect independer finish it within esis. nission of the t	ney compa ntly, toget a given pe	actly pr her wit riod of	esent and h the sup time.	in a d dis bervi	well- cuss their sor and
6	own wor perhaps <b>Examina</b> Graded r <b>Formalit</b> Successf Course a Seminar	ed manne k. They o an exterr tion module e y of exan ul pass of chieveme ("Oberse	er and discuss the design their proj nal partner, and xam: Master The nination f thesis. ent prior to subn	em critically. T ect independer finish it within esis. nission of the t	ney compa ntly, toget a given pe	actly pr her wit riod of	esent and h the sup time.	in a d dis bervi	well- cuss their sor and
5	own wor perhaps <b>Examina</b> Graded r <b>Formalit</b> Successf Course a Seminar <b>Module</b>	ed manne k. They o an exterr tion module e y of exan ul pass of chieveme ("Oberse requirem	er and discuss the design their proj- nal partner, and xam: Master The nination thesis. ent prior to subn minar"). nents (prerequis	em critically. T ect independer finish it within esis. nission of the t ites)	hey compa htly, toget a given pe	actly pr her wit riod of	esent and h the sup time.	in a d dis bervis	well- cuss their sor and
5	own wor perhaps Examina Graded r Formalit Successf Course a Seminar Module	ed manne k. They o an exterr tion module ex y of exan ul pass of chieveme ("Oberse requirem	er and discuss the design their proj- nal partner, and xam: Master The nination f thesis. ent prior to subn minar"). nents (prerequis efore October 20	em critically. T ect independen finish it within esis. nission of the t ites)	hey compa htly, toget a given pe	actly pr her wit riod of	esent and h the sup time.	in a d dis bervis	well- cuss their sor and
5	own wor perhaps <b>Examina</b> Graded r <b>Formalit</b> Successf Course a Seminar <b>Module</b> Start of s Learning	ed manne k. They o an exterr tion module e y of exan ul pass of chieveme ("Oberse requirem studies be ) and mo	er and discuss the design their proj- nal partner, and xam: Master The <b>nination</b> f thesis. ent prior to subn minar"). <b>nents (prerequis</b> efore October 20 dule MD 4 (Proje	em critically. T ect independen finish it within esis. nission of the t ites) 023: Successful ect Work).	hey compand htly, toget a given pe hesis: pres	actly pr her wit riod of sentatio	esent and h the sup time.	in a d dis bervis Adva	well- cuss their sor and anced ed Statistica
5	own wor perhaps <b>Examina</b> Graded r <b>Formalit</b> Successf Course a Seminar <b>Module</b> Start of s Learning Start of s	ed manne k. They o an exterr tion module en y of exan ul pass of chieveme ("Oberse requirem studies be ) and mo studies af	er and discuss the design their proj- nal partner, and xam: Master The nination f thesis. ent prior to subn minar"). nents (prerequis efore October 20 dule MD 4 (Proje ter October 202	em critically. T ect independen finish it within esis. nission of the t ites) 023: Successful ect Work). 3: At least 50 c	hey compand htly, toget a given pe hesis: pres pass of mo redit point	actly pr her wit riod of sentatio odule N ts, inclu	esent and h the sup time. On in the MD 1 (Adv uding 9 cr	in a d disc bervis Adva vanc	well- cuss their sor and anced ed Statistica points for
5	own wor perhaps <b>Examina</b> Graded r <b>Formalit</b> Successf Course a Seminar <b>Module</b> Start of s Learning Start of s Module	ed manne k. They o an exterr tion module en y of exan ul pass of chieveme ("Oberse requirem studies be ) and mo studies af 2 ("Statis	er and discuss the design their proj- nal partner, and xam: Master The nination thesis. ent prior to subn minar"). nents (prerequis efore October 20 dule MD 4 (Projecter October 202 tical Theory"), an	em critically. T ect independen finish it within esis. nission of the t ites) 023: Successful ect Work). 3: At least 50 c nd successfully	hey compand htly, toget a given pe hesis: pres pass of mo redit point	actly pr her wit riod of sentatio odule N ts, inclu	esent and h the sup time. On in the MD 1 (Adv uding 9 cr	in a d disc bervis Adva vanc	well- cuss their sor and anced ed Statistica points for
5	own wor perhaps <b>Examina</b> Graded r <b>Formalit</b> Successf Course a Seminar <b>Module</b> Start of s Learning Start of s Module "Case St	ed manne k. They of an exterr tion module en y of exan ul pass of chieveme ("Oberse requirem studies be ) and mo studies af 2 ("Statis udies" fro	er and discuss the design their proj- nal partner, and xam: Master The nination f thesis. ent prior to subner minar"). nents (prerequis dule MD 4 (Projecter October 202 dule MD 4 (Projecter October 202 tical Theory"), an om Module 4 ("P	em critically. T ect independen finish it within esis. nission of the t ites) 023: Successful ect Work). 3: At least 50 c nd successfully	hey compand htly, toget a given pe hesis: pres pass of mo redit point	actly pr her wit riod of sentatio odule N ts, inclu	esent and h the sup time. On in the MD 1 (Adv uding 9 cr	in a d disc bervis Adva vanc	well- cuss their sor and anced ed Statistica points for
7	own wor perhaps Examina Graded r Formalit Successf Course a Seminar Module Start of s Learning Start of s Module "Case Sti Allocatic	ed manne k. They of an extern tion module en y of exan ul pass of chieveme ("Oberse requirem studies be ) and mo studies af 2 ("Statis" udies" fro on to curr	er and discuss the design their proj- nal partner, and xam: Master The <b>nination</b> thesis. ent prior to subneminar"). <b>nents (prerequis</b> efore October 20 dule MD 4 (Projecter October 202 tical Theory"), an om Module 4 ("P <b>iculum</b>	em critically. T ect independen finish it within esis. nission of the t ites) 023: Successful ect Work). 3: At least 50 c nd successfully roject work").	hey compand htly, toget a given pe hesis: pres pass of me redit point completed	actly pr her wit riod of sentatio odule N ts, inclu d at lea	esent and h the sup time. On in the MD 1 (Adv uding 9 cr	in a d disc bervis Adva vanc	well- cuss their sor and anced ed Statistica points for
5 6 7 8	own wor perhaps Examina Graded r Formalit Successf Course a Seminar Module Start of s Learning Start of s Module "Case Str Allocatic Compuls	ed manne k. They of an exterr tion module extern y of exan ul pass of chieveme ("Oberse requirem studies be ) and mo studies af 2 ("Statis udies" fro on to curr ory modu	er and discuss the design their proj- nal partner, and xam: Master The nination f thesis. ent prior to subner minar"). nents (prerequis dule MD 4 (Projecter October 202 dule MD 4 (Projecter October 202 tical Theory"), an om Module 4 ("P	em critically. T ect independen finish it within esis. nission of the t ites) 023: Successful ect Work). 3: At least 50 c nd successfully roject work").	hey compand htly, toget a given pe hesis: pres pass of mo redit point complete Data Scie	actly pr her wit riod of sentatio odule N ts, inclu d at lea	esent and h the sup time. On in the MD 1 (Adv uding 9 cr	in a d disc bervis Adva vanc	well- cuss their sor and anced ed Statistica points for
6 7	own wor perhaps Examina Graded r Formalit Successf Course a Seminar Module Start of s Learning Start of s Module "Case Str Allocatic Compuls	ed manne k. They of an extern tion module en y of exan ul pass of chieveme ("Oberse requirem studies be ) and mo studies af 2 ("Statis udies" fro on to curr ory modu ibility	er and discuss the design their proj- nal partner, and xam: Master The <b>nination</b> thesis. ent prior to subneminar"). <b>nents (prerequis</b> efore October 20 dule MD 4 (Projecter October 202 tical Theory"), an om Module 4 ("P <b>iculum</b>	em critically. T ect independen finish it within esis. nission of the t ites) 023: Successful ect Work). 3: At least 50 c nd successfully roject work").	hey compand htly, toget a given pe hesis: pres pass of me redit point completed	actly pr her wit riod of sentatio odule N ts, inclu d at lea <u>nce</u> <b>nt</b>	esent and h the sup time. On in the MD 1 (Adv Iding 9 cr Ist the su	in a solution of the second se	well- cuss their sor and anced ed Statistica points for erformance

#### **Elective Courses**

In the entire elective area – "Methods" and "Applications" – modules with a total of 45 credit points are to be chosen. Only courses not already chosen in the qualifying bachelor's degree may be selected.

**Preliminary note:** Offered modules in this elective area are subject to regular fluctuations, and often instructors offer special courses on current research topics at short notice. Many courses in this elective area offered only on an irregular basis. Current information can always be found in the course catalogue of the respective semester. In order to take this into account, we do not provide a complete list of contents and competencies; these can be taken from the respective course announcements. Often, we import modules from other study programs, in these cases we refer to the original module handbooks. Students are only allowed to choose modules that have not been chosen otherwise, neither in the Master Data Science nor in their qualifying Bachelor program.

**General note on competences:** The students have deep knowledge about specialized methods of data science and related fields and apply them appropriately. They understand their theoretical background and choose suitable methods for a given problem. The students familiarize themselves with unknown fields of knowledge.

#### "Methods" Elective Courses

Modules with a total of at least 24 credit points are to be chosen.

At the request of the students, modules other than those mentioned here may be approved as elective modules via the chairman of the examination board.

Module MD E1-1: Time Series Analysis							
Rota	Duration	Semester	ECTS points	Workload			
as offered 1 semester 1st to 3rd 9 270							

1	Module structure							
	No.	Lecture/Course	Туре	ECTS	Hours			
	1	Time Series Analysis	L+T	9	4 + 2			
2	Language	2						
	English							
3	Module imported from							
	Statistics	department, specialized course for me	odules MS 6 / MS 7	in the M.Sc	. Statistics			
4	Examina	tion						
	Graded n	nodule exam.						
5	Formality	y of examination						
	Exam as a	announced by the lecturer(s).						
6	Module	requirements (prerequisites)						
	– none –							
7	Allocatio	n to curriculum						
	Elective r	nodule from the "Methods" catalogue	for M. Sc. study p	rogramme D	ata Science			
8	Responsi	bility	Department					
	Chairmar	n of board of examiners	Statistics					
			1 5					

Module MD E1-2: Survival Analysis							
Rota	Duration	Semester	ECTS points	Workload			
as offered 1 semester 1st to 3rd 9 270							

Module structure							
No.	Lecture/Course	Туре	ECTS	Hours			
1	Survival Analysis	L+T	9	4 + 2			
Language							
English							
Module i	mported from						
Statistics	department, specialized course for mo	odules MS 6 / MS	5 7 in the M.So	2. Statistics			
Examinat	ion						
Graded m	nodule exam.						
Formality	of examination						
Exam as a	announced by the lecturer(s).						
Module r	equirements (prerequisites)						
– none –							
Allocatio	n to curriculum						
Elective n	nodule from the "Methods" catalogue	for M. Sc. study	programme D	ata Science			
Responsi	bility	Department					
Prof. Dr. /	A. Groll, Prof. Dr. M. Pauly	Statistics					
	1 Language English Module i Statistics Examinat Graded m Formality Exam as a Module r – none – Allocatio Elective n Responsi	1       Survival Analysis         Language         English         Module imported from         Statistics department, specialized course for module         Examination         Graded module exam.         Formality of examination         Exam as announced by the lecturer(s).         Module requirements (prerequisites)         – none –         Allocation to curriculum         Elective module from the "Methods" catalogue         Responsibility	1       Survival Analysis       L + T         Language       English       L + T         Module imported from       Statistics department, specialized course for modules MS 6 / MS         Statistics department, specialized course for modules MS 6 / MS         Examination         Graded module exam.         Formality of examination         Exam as announced by the lecturer(s).         Module requirements (prerequisites)         – none –         Allocation to curriculum         Elective module from the "Methods" catalogue for M. Sc. study         Responsibility	1       Survival Analysis       L + T       9         Language       English       Module imported from         Statistics department, specialized course for modules MS 6 / MS 7 in the M.Sc         Examination         Graded module exam.         Formality of examination         Exam as announced by the lecturer(s).         Module requirements (prerequisites)         – none –         Allocation to curriculum         Elective module from the "Methods" catalogue for M. Sc. study programme D         Responsibility			

Module MD E1-3: B	ootstrapping						
Rota	Duration	Semester	ECTS points	Workload			
as offered 1 semester 1st to 3rd 9 270							

1	Module	structure						
	No.	Lecture/Course	Туре	ECTS	Hours			
	1	Bootstrapping	L+T	9	4 + 2			
2	Language							
	English							
3	Module i	mported from						
	Statistics	department, specialized course for r	modules MS 6 / N	1S 7 in the M.	Sc. Statistics			
4	Examina	tion						
	Graded n	nodule exam.						
5	Formality	y of examination						
	Exam as a	announced by the lecturer(s).						
6	Module I	requirements (prerequisites)						
	– none –							
7	Allocatio	n to curriculum						
	Elective r	module from the "Methods" catalogu	ie for M. Sc. stud	y programme	Data Science			
8	Responsi	bility	Department					
	Prof. Dr.	C. Jentsch, Prof. Dr. M. Pauly	Statistics					

Module MD E1-4: Stichprobenverfahren						
Rota	Duration	Semester	ECTS points	Workload		
as offered	1 semester	1st to 3rd	5	150		

1	Module s	structure			Module structure						
	No.	Lecture/Course	Туре	ECTS	Hours						
	1	Stichprobenverfahren	L + T	5	2 + 1						
2	Language	2									
	German										
3	Module i	mported from									
	Statistics	department, part of module MS 3 in	the M.Sc. Statisti	cs							
4	Examinat	tion									
	Graded n	nodule exam.									
5	Formality	y of examination									
	Exam as a	announced by the lecturer(s).									
6	Module r	equirements (prerequisites)									
	– none –										
7	Allocatio	n to curriculum									
	Elective r	nodule from the "Methods" catalogu	e for M. Sc. study	<sup>,</sup> programme	e Data Science						
8	Responsi	bility	Department								
	Prof. Dr.	Ph. Doebler, Prof. Dr. G. Knapp,	Statistics								
	Prof. Dr.	C. Müller									

Module MD E1-5: Nonparametric Methods					
Rota	Duration	Semester	ECTS points	Workload	
as offered	1 semester	1st to 3rd	4,5	135	

1	1 Module structure							
	No.	Lecture/Course	Туре	ECTS	Hours			
	1	Nonparametric Methods	L + T	4,5	2 + 1			
2	Language	2						
	English o	r German						
3	Module i	imported from						
	Statistics	department, specialized course for m	odules MS 6 / MS	7 in the M.	Sc. Statistics			
4	Examinat	tion						
	Graded n	nodule exam.						
5	Formality	y of examination						
	Exam as a	announced by the lecturer(s).						
6	Module I	requirements (prerequisites)						
	– none –							
7	Allocatio	n to curriculum						
	Elective r	module from the "Methods" catalogue	e for M. Sc. study	programme	Data Science			
8	Responsi	bility	Department					
	Prof. Dr.	C. Müller	Statistics					

Module MD E1-6: Robust Methods						
Rota	Duration	Semester	ECTS points	Workload		
as offered	1 semester	1st to 3rd	4,5	135		

1	Module structure						
	No.Lecture/CourseTypeECTSHours						
	1	Robust Methods	L+T	4,5	2 + 1		
2	Language	2					
	English o	r German					
3	Module i	mported from					
	Statistics	department, specialized course for m	odules MS 6 / MS 3	7 in the M.Sc	. Statistics		
4	Examinat	tion					
	Graded n	nodule exam.					
5	Formality	y of examination					
	Exam as a	announced by the lecturer(s).					
6	Module I	requirements (prerequisites)					
	– none –						
7	Allocatio	n to curriculum					
	Elective r	module from the "Methods" catalogue	e for M. Sc. study p	rogramme D	ata Science		
8	Responsi	bility	Department				
	Prof. Dr.	R. Fried, Prof. Dr. C. Müller	Statistics				

Module MD E1-7: Optimization					
Rota	Duration	Semester	ECTS points	Workload	
as offered	1 semester	1st to 3rd	5	150	

1	Module structure						
	No.Lecture/CourseTypeECTSHours						
	1	Optimization	L+T	5	2 + 1		
2	Language	9					
	English						
3	Module i	mported from					
	Mathema	atics department, export course for th	e module <u>AR-214</u>	from the M.S	Sc. Automation		
	and Robo	otics					
4	Examinat	tion					
	Graded n	nodule exam.					
5	Formality	y of examination					
	Exam as a	announced by the lecturer(s).					
6	Module ı	requirements (prerequisites)					
	– none –						
7	Allocatio	n to curriculum					
	Elective r	nodule from the "Methods" catalogue	e for M. Sc. study (	orogramme D	ata Science		
8	Responsi	bility	Department				
	Dean of S	Studies Mathematics	Mathematics				

#### Module MD E1-8: Numerical Solution of Differential Equations

Rota	Duration	Semester	ECTS points	Workload
as offered	1 semester	1st to 3rd	5	150

1	Module structure						
	No.	Lecture/Course	Туре	ECTS	Hours		
	1	Numerical Solution of Differential	L+T	5	2 + 1		
		Equations					
2	Language	2					
	English						
3	Module i	mported from					
	Mathema	atics department, export course for th	e module <u>AR-308</u>	from the M.S	c. Automation		
	and Robo	otics					
4	Examina	tion					
	Graded n	nodule exam.					
5	Formality	y of examination					
	Exam as	announced by the lecturer(s).					
6	Module	requirements (prerequisites)					
	– none –						
7	Allocatio	n to curriculum					
	Elective r	nodule from the "Methods" catalogue	e for M. Sc. study p	programme D	ata Science		
8	Responsi	bility	Department				
	Dean of S	Studies Mathematics	Mathematics				

Module MD E1-9: Generalized Linear Models					
Rota	Duration	Semester	ECTS points	Workload	
as offered	1 semester	1st to 3rd	9	270	

1	1 Module structure							
	No.	Lecture/Course	Туре	ECTS	Hours			
	1	Generalized Linear Models	L + T	9	4 + 2			
2	Languag	e						
	English							
3	Module	imported from						
	Statistics	department, specialized course ir	h the M.Sc. Statistics	5				
4	Examina	tion						
	Graded	nodule exam.						
5	Formalit	y of examination						
	Exam as	announced by the lecturer(s).						
6	Module	requirements (prerequisites)						
	– none –	-						
7	Allocatio	on to curriculum						
	Elective	module from the "Methods" catal	ogue for M. Sc. stud	y programme	e Data Science			
8	Respons	ibility	Department					
	Prof. Dr.	A. Groll	Statistics					

#### Module MD E1-10: Introduction to Computational Fluid Dynamics

Rota	Duration	Semester	ECTS points	Workload
as offered	1 semester	1st to 3rd	5	150

1	Module	Module structure					
	No.Lecture/CourseTypeECTSHours						
	1     Introduction to Computational Fluid     L + T     5     2 + 1						
	Dynamics						
2	Language						
	English						
3	Module imported from						
	Mathematics department, module MAT-708						
4	Examination						
	Graded module exam.						
5	Formality of examination						
	Exam as announced by the lecturer(s).						
6	Module requirements (prerequisites)						
	– none –						
7	Allocation to curriculum						
	Elective module from the "Methods" catalogue for M. Sc. study programme Data Science						
8	Responsi	bility	Department				
	Dean of S	Studies Mathematics	Mathematics				

Module MD E1-11: Introduction to Computational Intelligence					
uration	Semester	ECTS points	Workload		
semester	1st to 3rd	4.5	135		
l	ration	ration Semester	ration Semester ECTS points		

1	Module	structure					
	No. Lecture/Course Type ECTS Hours						
1Introduction to ComputationalL + T4.52 + 1							
	Intelligence						
2	Language	2					
	English	English					
3	Module imported from						
	Computer Science department, module <u>INF-Bsc-305</u>						
4	Examination						
	Graded module exam.						
5	Formality of examination						
	Exam as announced by the lecturer(s).						
6	Module requirements (prerequisites)						
	– none –						
7	Allocation to curriculum						
	Elective module from the "Methods" catalogue for M. Sc. study programme Data Science						
8	Responsi	bility	Department				
	Dean of S	Studies	Computer Scienc	е			
		·	20				

#### Module MD E1-12: Architecture and Implementation of Database Systems

Rota	Duration	Semester	ECTS points	Workload
as offered	1 semester	1st to 3rd	8	240

1	Module structure						
	No. Lecture/Course Type ECTS Hours						
	1Architecture and Implementation ofL + T84 + 2						
	Database Systems						
2	Language						
	English						
3	Module imported from						
	Computer Science department, module INF-MSc-214						
4	Examination						
	Graded module exam.						
5	Formality of examination						
	Exam as announced by the lecturer(s).						
6	Module requirements (prerequisites)						
	– none –						
7	Allocation to curriculum						
	Elective module from the "Methods" catalogue for M. Sc. study programme Data Science						
8	Responsi	bility	Department				
	Prof. Dr.	J. Teubner	Computer Scien	се			

Module MD E1-13: Cyl	per-Physical Systen	n Fundamentals		
Rota	Duration	Semester	ECTS points	Workload
as offered	1 semester	1st to 3rd	6	180

1	Module structure					
	No. Lecture/Course Type ECTS Hours					
	1	Cyber-Physical System Fundamental	s L+T	6	4 + 2	
2	Language					
	English					
3	Module imported from					
	Computer Science department, Mathematics department, export course for the module AR-					
	215 from the M.Sc. Automation and Robotics					
4	Examination					
	Graded module exam.					
5	Formality of examination					
	Exam as announced by the lecturer(s).					
6	Module requirements (prerequisites)					
	– none –					
7	Allocation to curriculum					
	Elective module from the "Methods" catalogue for M. Sc. study programme Data Science			Data Science		
8	Responsi	bility	Department			
	Prof. Dr.	JJ. Chen	<b>Computer Scien</b>	ce		

# Module MD E1-14: Real-Time Systems and ApplicationsRotaDurationSemesterECTS pointsWorkload

as offered 1 semester 1st to 3rd 8 240	Rota	Duration	Semester	ECTS points	Workload
	as offered	1 semester	1st to 3rd	8	240

1	Module structure				
	No.	Lecture/Course	Туре	ECTS	Hours
	1	Real-Time Systems and Applications	L+T	8	4 + 2
2	Language	2			
	English				
3	Module imported from				
	Computer Science department, module INF-MSc-223				
4	Examination				
	Graded module exam.				
5	Formality of examination				
	Exam as announced by the lecturer(s).				
6	Module requirements (prerequisites)				
	– none –				
7	Allocation to curriculum				
	Elective module from the "Methods" catalogue for M. Sc. study programme Data Science			oata Science	
8	Responsi	bility	Department		
	Prof. Dr.	JJ. Chen	Computer Scien	ce	

Module MD E1-15: Mo	odel Selection and	Model Averaging		
Rota	Duration	Semester	ECTS points	Workload
as offered	1 semester	1st to 3rd	9	270

1	Module	structure				
	No.	Lecture/Course	Туре	ECTS	Hours	
	1	Model Selection and Model Averagir	ng L+T	9	4 + 2	
2	Language	Language				
	English	English				
3	Module imported from					
	Statistics department, specialized course for modules MS 6 / MS 7 in the M.Sc. Statistics					
4	Examination					
	Graded module exam.					
5	Formality of examination					
	Exam as announced by the lecturer(s).					
6	Module requirements (prerequisites)					
	– none –					
7	Allocation to curriculum					
	Elective module from the "Methods" catalogue for M. Sc. study programme Data Science					
8	Responsi	ibility	Department			
	Dean of S	Studies	Statistics			

Module MD E1-16: Asymptotic Theory									
Ro	ta		Duration	Semester		ECTS	points	W	Vorkload
as	offered		1 semester	1st to 3rd		4,5		1	35
1	Module	structure							
	No.	Lecture	/Course		Туре		ECTS		Hours
	1	Asympt	otic Theory		L+T		4,5		2 + 1
2	Language	9							
	English								
3	Module imported from								
	Statistics	departm	ent, specialized o	course for modu	les MS 6	5 / MS 7	7 in the N	1.Sc	. Statistics
4	Examina	tion							

4	Examination	
	Graded module exam.	
5	Formality of examination	
	Exam as announced by the lecturer(s).	
6	Module requirements (prerequisites)	
	– none –	
7	Allocation to curriculum	
	Elective module from the "Methods" catalogue	e for M. Sc. study programme Data Science
8	Responsibility	Department
	Dean of Studies	Statistics

Module MD E1-17: Machine Learning Paradigms for Complex Data						
Rota	Duration	Semester	ECTS points	Workload		
as offered	1 semester	1st to 3rd	8	240		

1	Module	structure				
	No.	Lecture/Course	Туре	ECTS	Hours	
	1	Machine Learning Paradigms for	L + T	8	4 + 2	
		Complex Data				
2	Languag	e				
	English					
3	Module	imported from				
	Compute	er Science department, module <u>INF-N</u>	<u>ISc-236</u>			
4	Examina	tion				
	Graded r	nodule exam.				
5	Formalit	y of examination				
	Exam as	announced by the lecturer(s).				
6	Module	requirements (prerequisites)				
	– none –					
7	Allocatio	on to curriculum				
	Elective module from the "Methods" catalogue for M. Sc. study programme Data Science					
8	Respons	ibility	Department			
	Prof. Dr.	E. Müller	Computer Scie	nce		

#### Module MD E1-18: Neural networks for solving ordinary differential equations

Rota	Duration	Semester	ECTS points	Workload
as offered	1 semester	1st to 3rd	5	150

1	Module s	structure					
	No.	Lecture/Course	Туре	ECTS	Hours		
	1	Neural networks for solving ordinary	L+T	5	2 + 1		
		differential equations					
2	Language	2					
	English						
3	Module i	imported from					
	Mathema	atics department, useable for module	<u>MAT-439</u>				
4	Examinat	tion					
	Graded n	nodule exam.					
5	Formality	y of examination					
	Exam as a	announced by the lecturer(s).					
6	Module I	requirements (prerequisites)					
	– none –						
7	Allocation to curriculum						
	Elective module from the "Methods" catalogue for M. Sc. study programme Data Science						
8	Responsi	bility	Department				
	Dean of S	Studies Mathematics	Mathematics				

Module MD E1-19: Numerical Methods for PDEs						
Rota	Duration	Semester	ECTS points	Workload		
as offered	1 semester	1st to 3rd	5	150		

1	Module s	structure				
	No.	Lecture/Course	Туре	ECTS	Hours	
	1	Numerical Methods for PDEs	L+T	5	2 + 1	
2	Language	9				
	English					
3	Module i	mported from				
	Mathema	atics department, export course for the	e module <u>AR-30</u>	<mark>8</mark> or AR-214 f	rom the M.Sc.	
	Automat	ion and Robotics				
4	Examinat	tion				
	Graded n	nodule exam.				
5	Formality	y of examination				
	Exam as a	announced by the lecturer(s).				
6	Module r	requirements (prerequisites)				
	– none –					
7	Allocatio	n to curriculum				
	Elective module from the "Methods" catalogue for M. Sc. study programme Data Science					
8	Responsi	bility	Department			
	Dean of S	Studies Mathematics	Mathematics			

#### Module MD E1-20: Artificial Intelligence

Rota	Duration	Semester	ECTS points	Workload
as offered	1 semester	1st to 3rd	3	90

1	Module s	structure					
	No.	Lecture/Course	Туре	ECTS	Hours		
	1	Artificial Intelligence	L+T	3	2 + 1		
2	Language	2					
	English						
3	Module i	mported from					
	Institute	of Neuroinformatics, Ruhr University	Bochum				
4	Examina	tion					
	Graded n	nodule exam.					
5	Formality	y of examination					
	Exam as a	announced by the lecturer(s).					
6	Module I	requirements (prerequisites)					
	– none –						
7	Allocation to curriculum						
	Elective module from the "Methods" catalogue for M. Sc. study programme Data Science						
8	Responsi	bility	Department				
	Dean of S	Studies	Statistics				

Module MD E1-21: Deep Learning						
Rota	Duration	Semester	ECTS points	Workload		
as offered	1 semester	1st to 3rd	9	270		
					]	

1	Module s	structure						
	No.	Lecture/Course	Туре	ECTS	Hours			
	1	Deep Learning	L+T	9	4 + 2			
2	Language	2		-				
	English							
3	Module i	mported from						
	Statistics	department, specialized course for m	odules MS 6 / MS	7 in the M.So	c. Statistics			
4	Examina	tion						
	Graded n	nodule exam.						
5	Formality	y of examination						
	Exam as	announced by the lecturer(s).						
6	Module I	requirements (prerequisites)						
	– none –							
7	Allocatio	n to curriculum						
	Elective module from the "Methods" catalogue for M. Sc. study programme Data Science							
8	Responsi	bility	Department					
	Dean of S	Studies	Statistics					
			25					

#### Module MD E1-22: Unit Root and Cointegration Analysis

Rota	Duration	Semester	ECTS points	Workload
as offered	1 semester	1st to 3rd	9	270

1	Module structure						
	No.	Lecture/Course	Туре	ECTS	Hours		
	1	Unit Root and Cointegration Analysis	s L+T	9	4 + 2		
2	Language	9					
	English						
3	Module i	mported from					
	Statistics	department, specialized course for m	odules MS 6 / MS	7 in the M.Sc	. Statistics		
4	Examinat	tion					
	Graded n	nodule exam.					
5	Formality	y of examination					
	Exam as a	announced by the lecturer(s).					
6	Module requirements (prerequisites)						
	– none –						
7	Allocatio	n to curriculum					
	Elective module from the "Methods" catalogue for M. Sc. study programme Data Science						
8	Responsi	bility	Department				
	Dean of S	Studies	Statistics				

Module MD E1-23: Panel Data Econometrics					
Rota	Duration	Semester	ECTS points	Workload	
as offered	1 semester	1st to 3rd	9	270	

1	Module structure						
	No.	Lecture/Course	Туре	ECTS	Hours		
	1	Panel Data Econometrics	L+T	9	4 + 2		
2	Languag	e					
	English						
3	Module	imported from					
	Statistics	s department, specialized course f	or modules MS 6 / N	/IS 7 in the M	.Sc. Statistics		
4	Examina	Examination					
	Graded I	module exam.					
5	Formalit	y of examination					
	Exam as	announced by the lecturer(s).					
6	Module	Module requirements (prerequisites)					
	– none –	-					
7	Allocatio	on to curriculum					
	Elective	module from the "Methods" cata	ogue for M. Sc. stud	ly programme	e Data Science		
8	Respons	ibility	Department				
	Dean of	Studioc	Statistics				

# Module MD E1-24: Text as DataRotaDurationSemesterECTS pointsWorkloadas offered1 semester1st to 3rd4,5135

1	Module structure						
	No.	Lecture/Course	Туре	ECTS	Hours		
	1	Text as Data	L + T	4,5	2 + 1		
2	Language	2					
	English						
3	Module i	mported from					
	Statistics	department, specialized course for m	odules MS 6 / MS	7 in the M.S	c. Statistics		
4	Examina	tion					
	Graded n	nodule exam.					
5	Formality	y of examination					
	Exam as a	announced by the lecturer(s).					
6	Module I	requirements (prerequisites)					
	– none –						
7	Allocatio	n to curriculum					
	Elective r	nodule from the "Methods" catalogue	e for M. Sc. study	programme l	Data Science		
8	Responsi	bility	Department				
	Dean of S	Studies	Statistics				

Module MD E1-25: Statistical Network Analysis					
Rota	Duration	Semester	ECTS points	Workload	
as offered	1 semester	1st to 3rd	4,5	135	

1	Module structure						
	No.	Lecture/Course	Туре	ECTS	Hours		
	1	Statistical Network Analysis	L+T	4,5	2 + 1		
2	Language	2					
	English						
3	Module i	mported from					
	Statistics	department, specialized course for m	odules MS 6 / MS	7 in the M.S	c. Statistics		
4	Examina	tion					
	Graded n	nodule exam.					
5	Formality	y of examination					
	Exam as a	announced by the lecturer(s).					
6	Module ı	equirements (prerequisites)					
	– none –						
7	Allocatio	n to curriculum					
	Elective r	nodule from the "Methods" catalogue	e for M. Sc. study	programme	Data Science		
8	Responsi	bility	Department				
	Dean of S	studies	Statistics				

#### Module MD E1-26: Selected Topics in Data Science

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Rota	Duration	Semester	ECTS points	Workload
as offered	1 semester	1st to 3rd	6	180

1	Module structure						
	No.	Lecture/Course	Туре	ECTS	Hours		
	1	Selected Topics in Data Science	L+T	6	2 + 2		
2	Language	9					
	English						
3	Module i	mported from					
	Statistics	department, specialized course for m	odules MS 6 / MS 7	7 in the M.Sc	. Statistics		
4	Examinat	tion					
	Graded n	nodule exam.					
5	Formality	y of examination					
	Exam as a	announced by the lecturer(s).					
6	Module r	requirements (prerequisites)					
	– none –						
7	Allocatio	n to curriculum					
	Elective module from the "Methods" catalogue for M. Sc. study programme Data Science						
8	Responsi	bility	Department				
	Dean of S	Studies	Statistics				

Module MD E1-27: Advanced Bayesian Data Analysis					
Rota	Duration	Semester	ECTS points	Workload	
as offered	1 semester	1st to 3rd	6	180	

1	Module structure							
	No.	Lecture/Course	Туре	ECTS	Hours			
	1	Advanced Bayesian Data Analysis	L+T	4,5	2 + 2			
2	Language	e						
	English							
3	Module	imported from						
	Statistics	department, specialized course for n	nodules MS 6 / M	S 7 in the M	.Sc. Statistics			
4	Examina	tion						
	Graded r	nodule exam.						
5	Formalit	y of examination						
	Exam as	announced by the lecturer(s).						
6	Module	Module requirements (prerequisites)						
	– none –							
7	Allocatio	n to curriculum						
	Elective I	module from the "Methods" catalogu	e for M. Sc. study	/ programme	e Data Science			
8	Respons	ibility	Department					
	Dean of S	Studios	Statistics					

In addition to the above, many German language courses from the three departments can be chosen. Please note that sufficient German skill are required for all these courses. If you want to choose one of these courses, please get in touch with the Data Science examination board prior to participation. For example, the following courses are possible:

- from the M. Sc. study programme "Mathematik" of the Department of Mathematics:
  - o Module MAT-406: Numerik II
  - o Module MAT-419: Diskrete Optimierung
  - o Module MAT-424: Nichtlineare Optimierung
- from the M. Sc. study programme "Informatik" of the Department of Computer Science:
  - o Module INF-MSc-231: Praktische Optimierung
  - o Module INF-MSc-232: Mustererkennung
  - o Module INF-MSc-233: Graphische Datenverarbeitung
  - o Module INF-MSc-234: Commonsense Reasoning
  - o Module INF-MSc-241: Algorithmen und Datenstrukturen
  - o Module INF-MSc-401: Modellbildung, Simulation und Analyse
  - o Module INF-MSc-407: Verteilte Programmierung und numerische Algorithmen
  - o Module INF-MSc-501: Ausgewählte Kapitel der Computational Intelligence
  - o Module INF-MSc-502: Computer Vision
  - o Module INF-MSc-505: Geometrische Modellierung
  - o Module INF-MSc-506: Maschinelles Lernen
  - o Module INF-MSc-508: Schrifterkennung
  - o Module INF-MSc-509: Fortgeschrittene Themen der Wissensrepräsentation
  - o Module INF-MSc-510: IT-Management
  - o Module INF-MSc-514: Computational Omics
  - o Module INF-MSc-518: Digitalisierung von Fertigungsprozessen
  - o Module INF-MSc-605: Datenbanktheorie
  - o Module INF-MSc-606: Algorithmische Bioinformatik
  - o Module INF-MSc-607: Evolutionäre Algorithmen

Modules with a total of at least 16 credit points are to be chosen.

At the request of the students, modules other than those mentioned here may be approved as elective modules via the chairman of the examination board.

**Preliminary note:** Offered modules in this elective area are subject to regular fluctuations, and often instructors offer special courses on current research topics at short notice. Many courses in this elective area offered only on an irregular basis. Current information can always be found in the course catalogue of the respective semester.

In order to take this into account, we do not provide a complete list of contents and competencies; these can be taken from the respective course announcements. Often, we import modules from other study programs, in these cases we refer to the original module handbooks. Students are only allowed to choose modules that have not been chosen otherwise, neither in the Master Data Science nor in their qualifying Bachelor program.

**General note on competences:** The students have deep knowledge about specialized methods of data science and related fields and apply them appropriately. They understand their theoretical background and choose suitable methods for a given problem. They apply methods to real data by means of software.

#### Module MD E2-1: Bioinformatics

Rota	Duration	Semester	ECTS points	Workload
as offered	1 semester	2nd to 3rd	9	270

1	Module s	structure				
	No.	Lecture/Course	Туре	ECTS	Hours	
	1	Bioinformatics	L+T	9	4 + 2	
2	Language	9				
	English					
3	Module imported from					
	Statistics	department, specialized course for m	odules MS 6 / MS	57 in the M.	Sc. Statistics	
4	Examinat	tion				
	Graded n	nodule exam.				
5	Formality	y of examination				
	Exam as a	announced by the lecturer(s).				
6	Module I	requirements (prerequisites)				
	– none –					
7	Allocatio	n to curriculum				
	Elective module from the "Applications" catalogue for M. Sc. study programme Data Science					
8	Responsi	bility	Department			
	Prof. Dr.	J. Rahnenführer	Statistics			

#### Module MD E2-2: Toxicology

Rota	Duration	Semester	ECTS points	Workload
as offered	1 semester	2nd to 3rd	9	270

1	Module	structure				
	No.	Lecture/Course	Туре	ECTS	Hours	
	1	Toxicology	L+T	9	4 + 2	
2	Language	e				
	English					
3	Module imported from					
	Statistics	department, specialized course for m	nodules MS 6 / MS	7 in the M.So	c. Statistics	
5	Examina	tion				
	Graded r	nodule exam.				
6	Formalit	y of examination				
	Exam as	announced by the lecturer(s).				
7	Module	requirements (prerequisites)				
	– none –					
8	Allocatio	n to curriculum				
	Elective module from the "Applications" catalogue for M. Sc. study programme Data Science					
9	Respons	ibility	Department			
	Chairmai	n of board of examiners	Statistics			

Module MD E2-3: Eco	nometrics				
Rota	Duration	Semester	ECTS points	Workload	
as offered	1 semester	2nd to 3rd	9	270	

1	Module structure						
	No.	Lecture/Course	Туре	ECTS	Hours		
	1	Econometrics	L+T	9	4 + 2		
2	Language	9					
	English						
3	Module i	mported from					
	Statistics	department, specialized course for r	nodules MS 6 / N	1S 7 in the M.	Sc. Statistics		
4	Examina	tion					
	Graded n	nodule exam.					
5	Formality	y of examination					
	Exam as a	announced by the lecturer(s).					
6	Module	requirements (prerequisites)					
	– none –						
7	Allocatio	n to curriculum					
	Elective module from the "Applications" catalogue for M. Sc. study programme Data Science						
8	Responsi	bility	Department				
	JProf. Dr.	A. Arsova, Prof. Dr. C. Jentsch	Statistics				

#### Module MD E2-4: Econometrics of treatment effects and policy evaluation

Rota	Duration	Semester	ECTS points	Workload
as offered	1 semester	2nd to 3rd	4,5	135

1	Module structure					
	No.	Lecture/Course	Туре	ECTS	Hours	
	1	Econometrics of treatment effects ar	nd L+T	4,5	2 + 1	
		policy evaluation				
2	Language	2				
	English					
3	Module i	mported from				
	Statistics	department, specialized course for m	odules MS 6 / MS	7 in the M.So	. Statistics	
4	Examina	tion				
	Graded n	nodule exam.				
5	Formality	y of examination				
	Exam as	announced by the lecturer(s).				
6	Module I	requirements (prerequisites)				
	– none –					
7	Allocation to curriculum					
	Elective module from the "Applications" catalogue for M. Sc. study programme Data Science					
8	Responsi	bility	Department			
	Prof. Dr.	C. Jentsch	Statistics			

Module MD E2-5: Natürlichsprachige Systeme						
Rota	Duration	Semester	ECTS points	Workload		
as offered	1 semester	2nd to 3rd	7	180		

1	Module	structure				
	No.	Lecture/Course	Туре	ECTS	Hours	
	1	Natürlichsprachige Systeme	L + T	7	2 + 2	
2	Language	2				
	German					
3	Module i	imported from				
	Compute	er Science department, module <u>INF-I</u>	<u> MSc-507</u>			
4	Examina	tion				
	Graded n	nodule exam.				
5	Formality	y of examination				
	Exam as	announced by the lecturer(s).				
6	Module	requirements (prerequisites)				
	– none –					
7	Allocatio	n to curriculum				
	Elective module from the "Applications" catalogue for M. Sc. study programme Data Science					
8	Responsi	ibility	Department			
	Prof. Dr.	K. Morik, Prof Dr. E. Schubert	Computer Scie	nce		

Module MD E2-6: Control Theory and Applications						
Rota	Duration	Semester	ECTS points	Workload		
as offered	1 semester	2nd to 3rd	7	210		

L L	would	Structure					
	No.	Lecture/Course	Туре	ECTS	Hours		
	1	Control Theory and Applications	L + T	7	3 + 2		
2	Language	e					
	English						
3	Module imported from						
	Departm	ent of Biochemical and Chemical Engi	ineering, manda	tory course fo	or the module		
	<u>AR-102</u> ir	n the M.Sc. Automation and Robotics					
4	Examina	tion					
	Graded n	nodule exam.					
5	Formality	y of examination					
	Exam as	announced by the lecturer(s).					
6	Module	requirements (prerequisites)					
	– none –						
7	Allocatio	n to curriculum					
	Elective module from the "Applications" catalogue for M. Sc. study programme Data Science						
8	Responsi	ibility	Department				
	Prof. Dr.	S. Engell	Biochemical ar	d Chemical E	ngineering		

Module MD E2-7: Reliability and Material Fatigue							
Rota	Duration	Semester	ECTS points	Workload			
as offered	1 semester	2nd to 3rd	9	270			

1	Module s	structure				
	No.	Lecture/Course	Туре	ECTS	Hours	
	1	Reliability and Material Fatigue	L + T	9	4 + 2	
2	Language	9				
	English or German					
3	Module imported from					
	Statistics	department, specialized course for m	odules MS 6 / MS	7 in the M.Sc	. Statistics	
4	Examinat	tion				
	Graded n	nodule exam.				
5	Formality	y of examination				
	Exam as a	announced by the lecturer(s).				
6	Module ı	requirements (prerequisites)				
	– none –					
7	Allocatio	n to curriculum				
	Elective module from the "Applications" catalogue for M. Sc. study programme Data Science					
8	Responsi	bility	Department			
	Prof. Dr.	C. Müller	Statistics			

Ro	ta		Duration	Semester		ECTS points	Workload		
as	offered		1 semester	2nd to 3r	d	9	270		
1	Module	structure							
	No.	Lecture	/Course		Туре	ECTS	Hours		
	1	Quality	Control		L+T	9	4 + 2		
2	Languag	ge							
	English	or Germar	1						
3	Module	imported	from						
		•	ent, specialized	course for mo	odules MS 6	5 / MS 7 in the N	A.Sc. Statistics		
4	-	Examination							
	Graded	module ex	am.						
5	Formali	ty of exam	ination						
	Exam as	announce	ed by the lecture	er(s).					
6	Module	requirem	ents (prerequisi	tes)					
	– none -	-							
7	Allocati	on to curri	iculum						
	Elective	module fr	om the "Applica	tions" catalo	gue for M. S	Sc. study progra	mme Data Scien		
8	Respons	sibility			Departmer	nt			
		C Müller			Statictics				

Prof. Dr. C. Müller	Statistics

Module MD E2-9: Financial Econometrics					
Rota	Duration	Semester	ECTS points	Workload	
as offered	1 semester	1st to 3rd	4,5	135	

Module structure					
No.	Lecture/Course	Туре	ECTS	Hours	
1	Financial Econometrics	L+T	4,5	2 + 1	
Language	2				
English					
Module i	mported from				
Statistics	department, specialized course for m	odules MS 6 / MS 3	7 in the M.Sc	. Statistics	
Examinat	tion				
Graded n	nodule exam.				
Formality	y of examination				
Exam as a	announced by the lecturer(s).				
Module r	equirements (prerequisites)				
– none –					
Allocation to curriculum					
Elective module from the "Methods" catalogue for M. Sc. study programme Data Science					
Responsi	bility	Department			
Dean of S	itudies	Statistics			
	No. 1 Language English Module i Statistics Examinat Graded n Formality Exam as a Module r – none – Allocatio Elective r Responsi	No.Lecture/Course1Financial EconometricsLanguageEnglishModule imported fromStatistics department, specialized course for mExaminationGraded module exam.Formality of examinationExam as announced by the lecturer(s).Module requirements (prerequisites)– none –Allocation to curriculum	No.       Lecture/Course       Type         1       Financial Econometrics       L + T         Language       English       L + T         Module imported from       Statistics department, specialized course for modules MS 6 / MS 7         Statistics department, specialized course for modules MS 6 / MS 7         Examination       Graded module exam.         Formality of examination       Exam as announced by the lecturer(s).         Module requirements (prerequisites)       – none –         – none –       –         Allocation to curriculum       Elective module from the "Methods" catalogue for M. Sc. study p         Responsibility       Department	No.       Lecture/Course       Type       ECTS         1       Financial Econometrics       L + T       4,5         Language       English       L + T       4,5         Module imported from       Statistics department, specialized course for modules MS 6 / MS 7 in the M.Sc         Statistics department, specialized course for modules MS 6 / MS 7 in the M.Sc       Examination         Graded module exam.       Formality of examination       Exam as announced by the lecturer(s).         Module requirements (prerequisites)       – none –       Allocation to curriculum       Elective module from the "Methods" catalogue for M. Sc. study programme D         Responsibility       Department	

#### Module MD E2-10: Machine Learning for Economic Data

Rota	Duration	Semester	ECTS points	Workload
as offered	1 semester	1st to 3rd	4,5	135

1	Module s	structure						
	No.	Lecture/Course	Туре	ECTS	Hours			
	1	Machine Learning for Economic Data	L+T	4,5	2 + 1			
2	Language	2						
	English							
3	Module imported from							
	Statistics	department, specialized course for m	odules MS 6 / N	IS 7 in the M	Sc. Statistics			
4	Examinat	tion						
	Graded n	nodule exam.						
5	Formality	y of examination						
	Exam as a	announced by the lecturer(s).						
6	Module ı	requirements (prerequisites)						
	– none –							
7	Allocatio	n to curriculum						
	Elective module from the "Methods" catalogue for M. Sc. study programme Data Science							
8	Responsi	bility	Department					
	Dean of S	Studies	Statistics					

Module MD E2-11: Bayesian Econometrics					
Rota	Duration	Semester	ECTS points	Workload	
as offered	1 semester	1st to 3rd	4,5	135	

1	Module s	Module structure					
	No.	Lecture/Course	Туре	ECTS	Hours		
	1	Bayesian Econometrics	L+T	4,5	2 + 1		
2	Language	2					
	English						
3	Module i	mported from					
	Statistics	department, specialized course for m	odules MS 6 / MS 7	7 in the M.Sc	. Statistics		
4	Examinat	tion					
	Graded n	nodule exam.					
5	Formality	y of examination					
	Exam as a	announced by the lecturer(s).					
6	Module ı	requirements (prerequisites)					
	– none –						
7	Allocatio	n to curriculum					
	Elective module from the "Methods" catalogue for M. Sc. study programme Data Science						
8	Responsi	bility	Department				
	Dean of S	Studies	Statistics				

#### Module MD E2-12: Statistics in Toxicology (Modelling)

Rota	Duration	Semester	ECTS points	Workload
as offered	1 semester	1st to 3rd	4,5	135

1	Module structure					
	No.	Lecture/Course	Туре	ECTS	Hours	
	1	Statistics in Toxicology (Modelling)	L+T	4,5	2 + 1	
2	Languag	e				
	English					
3	Module	imported from				
	Statistics	department, specialized course for m	odules MS 6 / N	1S 7 in the M	Sc. Statistics	
4	Examina	tion				
	Graded r	nodule exam.				
5	Formalit	y of examination				
	Exam as	announced by the lecturer(s).				
6	Module	requirements (prerequisites)				
	– none –					
7	Allocatio	n to curriculum				
	Elective I	module from the "Methods" catalogu	e for M. Sc. stud	y programme	e Data Science	
8	Respons	ibility	Department			
	Dean of s	Studies	Statistics			

Module MD E2-13: Statistics in Toxicology (Testing)					
Rota	Duration	Semester	ECTS points	Workload	
as offered	1 semester	1st to 3rd	4,5	135	

1	Module structure					
	No.	Lecture/Course	Туре	ECTS	Hours	
	1	Statistics in Toxicology (Testing)	L + T	4,5	2 + 1	
2	Language					
	English	English				
3	Module imported from					
	Statistics department, specialized course for modules MS 6 / MS 7 in the M.Sc. Statistics					
4	Examination					
	Graded module exam.					
5	Formality of examination					
	Exam as announced by the lecturer(s).					
6	Module requirements (prerequisites)					
	– none –					
7	Allocation to curriculum					
	Elective module from the "Methods" catalogue for M. Sc. study programme Data Science					
8	Responsi	bility	Department			
	Dean of S	Studies	Statistics			

#### Module MD E2-14: Statistics in Genetics (Bioinformatics)

Rota	Duration	Semester	ECTS points	Workload
as offered	1 semester	1st to 3rd	9	270

1	Module structure						
	No.	Lecture/Course	Туре	ECTS	Hours		
	1	Statistics in Genetics (Bioinformatics)	L+T	9	4 + 2		
2	Languag	ge					
	English						
3	Module	imported from					
	Statistic	s department, specialized course for mo	odules MS 6 / N	/IS 7 in the M	Sc. Statistics		
4	Examina	ation					
	Graded	module exam.					
5	Formali	ty of examination					
	Exam as	announced by the lecturer(s).					
6	Module	requirements (prerequisites)					
	– none -	-					
7	Allocati	on to curriculum					
	Elective	module from the "Methods" catalogue	for M. Sc. stud	y programme	e Data Science		
8	Respon	sibility	Department				
	Dean of	Studies	Statistics				

# Module MD E2-15: Introduction to Linguistic Data Science with RRotaDurationSemesterECTS pointsWorkloadas offered1 semester1st to 3rd5150

1	Module structure						
	No.	Lecture/Course	Туре	ECTS	Hours		
	1	Introduction to Linguistic Data Scienc	e L+T	5	2 + 1		
		with R					
2	Language	2					
	English						
3	Module i	imported from					
	Linguistic	c Data Science working group, Ruhr Uni	iversity Bochun	n			
4	Examina	tion					
	Graded n	nodule exam.					
5	Formality	y of examination					
	Exam as	announced by the lecturer(s).					
6	Module	requirements (prerequisites)					
	– none –						
7	Allocatio	n to curriculum					
	Elective r	module from the "Methods" catalogue	for M. Sc. stud	y programme	e Data Science		
8	Responsi	bility	Department				
	Dean of S	Studies	Statistics				

## Module MD E2-16: Introduction to Computational Linguistics with Python

Rota	Duration	Semester	ECTS points	Workload
as offered	1 semester	1st to 3rd	5	150

1	Module	structure				
	No.	Lecture/Course	Туре	ECTS	Hours	
	1	Introduction to Computational	L+T	5	2 + 1	
		Linguistics with Python				
2	Language	e				
	English					
3	Module	imported from				
	Linguistic	c Data Science working group, Ruhr Ur	niversity Bochun	n		
4	Examina	tion				
	Graded r	nodule exam.				
5	Formalit	y of examination				
	Exam as	announced by the lecturer(s).				
6	Module	requirements (prerequisites)				
	– none –					
7	Allocatio	n to curriculum				
	Elective module from the "Methods" catalogue for M. Sc. study programme Data Science					
8	Respons	ibility	Department			
	Dean of S	Studies	Statistics			

Module MD E2-17: Statistics meets Linguistics					
Rota	Duration	Semester	ECTS points	Workload	
as offered	1 semester	1st to 3rd	4,5	135	

1	Module s	Module structure					
	No.	Lecture/Course	Туре	ECTS	Hours		
	1	Statistics meets Linguistics	L+T	4,5	2 + 1		
2	Language	2					
	English						
3	Module i	mported from					
	Applied I	inguistics working group, TU Dortmun	d University				
4	Examinat	tion					
	Graded n	nodule exam.					
5	Formality	y of examination					
	Exam as a	announced by the lecturer(s).					
6	Module ı	requirements (prerequisites)					
	– none –						
7	Allocatio	n to curriculum					
	Elective r	nodule from the "Methods" catalogue	e for M. Sc. study	programme D	ata Science		
8	Responsi	bility	Department				
	Dean of S	Studies	Statistics				

## Module MD E2-18: Research Methods in English Linguistics

Rota	Duration	Semester	ECTS points	Workload
as offered	1 semester	1st to 3rd	4,5	135

1	Module s	structure			
	No.	Lecture/Course	Туре	ECTS	Hours
	1	Research Methods in English Linguisti	cs L+T	4,5	2 + 1
2	Language	2			
	English				
3	Module i	mported from			
	Applied li	inguistics working group, TU Dortmund	University		
4	Examinat	tion			
	Graded m	nodule exam.			
5	Formality	y of examination			
	Exam as a	announced by the lecturer(s).			
6	Module r	requirements (prerequisites)			
	– none –				
7	Allocatio	n to curriculum			
	Elective r	nodule from the "Methods" catalogue	for M. Sc. stu	idy programme	Data Science
8	Responsi	bility	Department		
	Dean of S	Studies	Statistics		

Module MD E2-19: Learning in Robotics						
Rota	Duration	Semester	ECTS points	Workload		
as offered	1 semester	1st to 3rd	4.5	150		

1	Module	Module structure					
	No.	Lecture/Course	Туре	ECTS	Hours		
	1	Learning in Robotics	L+T	4.5	2 + 1		
2	Language	2					
	English						
3	Module i	mported from					
	Departm	ent Electrical Engineering & Informati	on Technology, ı	nodule <u>AR-3</u> 2	<u>10</u> from the		
	M.Sc. Au	tomation and Robotics					
4	Examina	tion					
	Graded n	nodule exam.					
5	Formality	y of examination					
	Exam as	announced by the lecturer(s).					
6	Module I	requirements (prerequisites)					
	– none –						
7	Allocatio	n to curriculum					
	Elective module from the "Methods" catalogue for M. Sc. study programme Data Science						
8	Responsi	bility	Department				
	Dean of S	Studies	Statistics				

#### Module MD E2-20: Data-Based Dynamic Modeling

Rota	Duration	Semester	ECTS points	Workload
as offered	1 semester	1st to 3rd	3	90

1	Module s	structure			
	No.	Lecture/Course	Туре	ECTS	Hours
	1	Data-Based Dynamic Modeling	L+T	3	2 + 1
2	Language	2			
	English				
3	Module i	mported from			
	Departm	ent of Biochemical and Chemical Engi	neering, module <u>A</u>	<u>R-206</u> from t	he M.Sc.
	Automat	ion and Robotics			
4	Examinat	tion			
	Graded n	nodule exam.			
5	Formality	y of examination			
	Exam as a	announced by the lecturer(s).			
6	Module r	requirements (prerequisites)			
	– none –				
7	Allocatio	n to curriculum			
	Elective r	nodule from the "Methods" catalogue	e for M. Sc. study p	rogramme D	ata Science
8	Responsi	bility	Department		
	Dean of S	Studies	Statistics		

Module MD E2-21: Industrial Data Science I					
Rota	Duration	Semester	ECTS points	Workload	
as offered	1 semester	1st to 3rd	5	150	

1	Module structure					
	No.	Lecture/Course	Туре	ECTS	Hours	
	1	Industrial Data Science I	L+T	5	2 + 1	
2	Language	9				
	English					
3	Module i	imported from				
	Statistics	department, specialized course for m	odules MS 6 / M	S 7 in the M.	Sc. Statistics	
4	Examina	tion				
	Graded n	nodule exam.				
5	Formality	y of examination				
	Exam as	announced by the lecturer(s).				
6	Module	requirements (prerequisites)				
	– none –					
7	Allocatio	n to curriculum				
	Elective r	nodule from the "Methods" catalogu	e for M. Sc. study	/ programme	e Data Science	
8	Responsi	bility	Department			
	Dean of S	Studies	Statistics			

Module MD E2-22: Ga	Module MD E2-22: Game Theory					
Rota	Duration	Semester	ECTS points	Workload		
as offered	1 semester	1st to 3rd	7.5	225		

1	Module	structure						
	No.	Lecture/Course	Туре	ECTS	Hours			
	1	Game Theory	L+T	7.5	4			
2	Language	Language						
	English							
3	Module i	imported from						
	Economi	<u>cs Department</u>						
4	Examina	tion						
	Graded n	nodule exam.						
5	Formality	y of examination						
	Exam as	announced by the lecturer(s).						
6	Module	requirements (prerequisites)						
	– none –							
7	Allocatio	n to curriculum						
	Elective r	module from the "Methods" catalogue	e for M. Sc. stud	y programme	e Data Science			
8	Responsi	ibility	Department					
	Dean of S	Studies	Department of	Business and	l Economics			

## Module MD E2-23: Introduction to Linguistic Data ScienceRota<br/>as offeredDuration<br/>2 semestersSemester<br/>1st to 3rdECTS points<br/>10Workload<br/>300

1	Module	structure			
	No.	Lecture/Course	Туре	ECTS	Hours
	1	Introduction to Linguistic Data Science	e L+T	10	2 + 1
					(2 semesters)
2	Language	2			
	English				
3	Module i	mported from			
	Linguistic	: Data Science working group, Ruhr Uni	versity Bochun	n, comprised	of "Introduction
	to Linguis	stic Models (with R)" and "Linguistic Da	ta Science"		
4	Examina	tion			
	Graded n	nodule exam.			
5	Formality	y of examination			
	Exam as a	announced by the lecturer(s).			
6	Module I	requirements (prerequisites)			
	– none –				
7	Allocatio	n to curriculum			
	Elective r	nodule from the "Methods" catalogue	for M. Sc. stud	y programme	Data Science
8	Responsi	bility	Department		
	Dean of S	Studies	Statistics		

### Module MD E2-24: Introduction to Computational Linguistics

Rota	Duration	Semester	ECTS points	Workload
as offered	2 semesters	1st to 3rd	10	300

1 Module structure					
	No.	Lecture/Course	Туре	ECTS	Hours
	1	Introduction to Computational	L+T	10	2 + 1
		Linguistics			(2 semesters)
2	Language	2			
	English				
3	Module i	mported from			
	Linguistic	: Data Science working group, Ruhr Ur	niversity Bochum, o	comprised of	"Introduction
	to Comp	utational Linguistics (with Python)" an	d "Computational	Linguistics ar	nd AI"
4	Examina	tion			
	Graded n	nodule exam.			
5	Formality	y of examination			
	Exam as	announced by the lecturer(s).			
6	Module I	requirements (prerequisites)			
	– none –				
7	Allocatio	n to curriculum			
	Elective module from the "Methods" catalogue for M. Sc. study programme Data Science				
8	Responsi	bility	Department		
	Dean of S	Studies	Statistics		

Module MD E2-25: Resampling & Simulations						
Rota	Duration	Semester	ECTS points	Workload		
as offered	1 semester	1st to 3rd	4,5	135		

1 Module structure							
	No.	Lecture/Course	Туре	ECTS	Hours		
	1	<b>Resampling &amp; Simulations</b>	L+T	4,5	2+1		
2	Languag	;e					
	English						
3	Module	imported from					
	Statistic	s department, specialized course for	modules MS 6 / N	1S 7 in the M	.Sc. Statistics		
4	Examina	ation					
	Graded	module exam.					
5	Formali	ty of examination					
	Exam as	announced by the lecturer(s).					
6	Module	Module requirements (prerequisites)					
	– none -	-					
7	Allocati	on to curriculum					
	Elective	module from the "Methods" catalog	ue for M. Sc. stud	y programme	e Data Science		
8	Respons	sibility	Department				
	Dean of Studies Statistics						

#### Possible requirements in case of conditional admission

In the case of conditional admission, some of the following additional requirements are usually assigned as conditions. However, this list is not exhaustive, it is for your information only, it is possible that conditions beyond this list may also be imposed.

#### Module MD Req1: Advanced Mathematics

				-
Rota	Duration	Semester	ECTS points	Workload
Winter semester,	1 semester	beginning of	7	210
annual		programme		

1	Module structure								
	No.	Lecture/Course	Туре	ECTS	Hours				
	1	Advanced Engineering Mathematics	L+T	7	3 + 2				
2	Languag	ge							
	English								
3	Content	:							
	• L	inear Algebra: Vector spaces, matrices a	and equation s	ystems, linea	r maps, Jordan-,				
	L L	U-, QR-, and singular value decompositi	on, numerical	aspects.					
		Differential Equation: Linear systems, dif coefficients.	ferential equa	tions with co	nstant				
	• l	aplace-Transform: Definition, convolution	on and applica	tion to differe	ential equations.				
	• [	Differential Calculus with several variable	es: Derivatives	, inverse and	implicit				
	f	unctions, Taylor expansion and extreme	values.						
	• 9	• Stability of Differential Equations: Theorems of Ljapunov and Poincaré-Ljapunov.							
	• \	Variational Calculus.							
	Literature:								
	•	Bajpai, Avinash C., Mathematics for engineers and scientists							
	•	Meyer, R.M., Essential mathematics for applied fields							
	•	<ul> <li>Lancaster, P., Tismenetsky, M., The theory of matrices</li> </ul>							
	•	<ul> <li>Lang, S., Linear algebra</li> </ul>							
	Slides								
4	Compet	Competences / Qualification Goals							
	The students are acquainted with fundamental mathematical techniques and thus prepared								
	for their future courses. They understand the underlying mathematical structures in their								
	field.								
5	Examina	ation							
	Module exam.								
6		ty of examination							
		exam (2 hours).							
7	Module	requirements (prerequisites)							
	- none -								
8		on to curriculum							
		requirement in case of conditional adm	ission to the N	/I. Sc. study pi	rogramme Data				
	Science								

## Module MD Req1: Advanced Mathematics 9 Responsibility Department Dean of Studies Mathematics Mathematics

Module MD Req2: Data Structures and Programming						
Rota every semester	<b>Duration</b> 1 semester	Semester beginning of	<b>ECTS points</b> 10	Workload 300		
		programme				

1	Module	structure								
	No.	Lecture/Course	Туре	ECTS	Hours					
	1	Reading Course Data Structures and	reading	10						
		Programming	course							
2	Langua	ge								
	English									
3	Conten	t								
	•	Programming Languages: introduction to	Java; concepts	of structure	d and object-					
		oriented programming.								
	•	Algorithms: sorting and searching on lists	, trees and grap	ohs.						
	•	Object oriented software: classes; secret	principle and e	ncapsulation	; message					
		exchange; inheritance; hierarchies; excep	tion handling; ន្	genericity; ol	oject-oriented					
		design.								
	•	Programming in Java.								
	Literatu	ire:								
		<ul> <li>James T. Streib, Takako Soma: Guide 1</li> </ul>	o Data Structu	res: A Concis	e Introduction					
		Using Java. Springer 2017.								
		<ul> <li>Takako Soma, James T. Streib: Guide 1</li> </ul>	o Java: A Conci	se Introduct	ion to					
		Programming. Springer 2014.								
	In addit	ion to the English language Reading Cour	se, students ca	n attend the	German					
	languag	ge lecture Datenstrukturen, Algorithmen u	ind Programmi	eren I (see m	odule BD 3 from					
	the Bac	helor Data Science). However, although t	nese courses ar	e similar in c	ontent, they are					
	not equ	ivalent, and reading the literature is man	datory even if t	he lecture is	attended.					
4	Compe	Competences / Qualification Goals								
	Student	ts deeply understand the informal basics t	or the descript	ion of progra	amming					
	languag	ses and their implementation within the f	ramework of th	e Java progr	amming					
	languag	e. Students handle the basics of object-o	riented prograr	n design, ind	ependently					
	formula	ate solution algorithms for given problems	and implemer	nt them as Ja	va programs.					
	They de	eply understand selected designs for obj	ect-oriented so	ftware const	ruction and					
	evaluat	e their usability.								
5	Examin	ation								
	Module	exam								
6	Formali	ity of examination								
	Digital h	nomework: A set of programming tasks ha	as to be solved	at home and	handed in within					
	a givon	time limit. Details are published at the be	ginning of each	evaminatio	n pariod					

7	Module requirements (prerequisites)					
	- none –					
8	Allocation to curriculum					
	Possible requirement in case of conditional ad	nission to the M. Sc. study programme Data				
	Science					
9	Responsibility	Department				
	Dean of Studies; Dr. S. Dissmann	Computer Science				
-	·	·				

Module MD Req3: Information Systems							
Rota	Duration	Semester	ECTS points	Workload			
every semester	1 semester	beginning of	5	150			
		programme					

1	Module structure								
	No.	Lecture/Course	Туре	ECTS	Hours				
	1a	Reading Course Information Systems	s reading	5					
			course						
	1b	Informationssysteme (See BD 13.1,	L+T	5	2 + 1				
		Bachelor Data Science; in German)							
2	Language	2							
	English (1	la) or German (1b)							
3	Content								
	One can	choose to attend either the English la	nguage reading c	ourse or the	German				
	language	lecture. Both cover the same content	ts:						
	The cour	se deals with the architecture and use	e of information s	ystems, esp	ecially database				
	and infor	mation retrieval systems. Powerful, o	leclarative query	and change	languages are				
		ack to computer-based, procedural ex	-	-					
	formalisa	ation of applications as well as the pra	ctical handling of	an object-r	elational				
		system (ORACLE) is explained.							
	Literatur	-							
		naterials, and some references given		urse "Inforn	nationssysteme"				
		Teubner of the Faculty of Computer So	cience						
4	-	ences / Qualification Goals							
		understand the basics of syntax and s							
		e architecture of information systems,	and perform the	developme	nt cycle of				
		ons based on this.							
5	Examina								
_		n for (1a), written exam in German la	nguage for (1b)						
6		y of examination							
	Examinat	tion based on the lecture materials.							
7	Module I	requirements (prerequisites)							
	- none -								
8	Allocatio	n to curriculum							
		requirement in case of conditional ad	mission to the M	. Sc. study p	rogramme Data				
	Science								
9	Responsi	-	Department						
	Prof. Dr.	J. Teubner	Computer Scien	ce					

Module MD Req4: Probability						
Rota	Duration	Semester	ECTS points	Workload		
every semester	1 semester	beginning of	5	150		
		programme				

1	Module	structure			
	No.	Lecture/Course	Туре	ECTS	Hours
	1	Reading Course Probability	reading	5	
			course		
2	Languag	e			
	English				
3	Content				
	В • S	oncepts of probability, distributions, ayes' rule, sequences of events. ampling, Binomial distribution, Norm	al approximation		
		andom variables, expectation and va			
		robability densities, Exponential and	Gamma distributi	ons, substitu	itions,
	-	umulative distribution functions.			
		pint distributions, Uniform and Norm			
		ependence, conditional distributions	, covariance and (	correlation.	
	Literatur				
		an: Probability. Springer 1993: Chapt	ers 1, 2.1, 2.2, 2.5	, 3.1-3.5, 4.1	., 4.2, 4.4, 4.5,
4	5.1-5.3,6	o. ences / Qualification Goals			
4	-	gain a deep understanding of proba	aility. They inden	ndontly into	arato statistical
		s in the context of probability theory			-
	-	apply mathematical proof technique			late methods.
5	Examina				
	Module				
	in outlie				
6	Formalit	y of examination			
	Examina	tion based on the book chapters.			
7	Module	requirements (prerequisites)			
	- none -				
8		on to curriculum			
		requirement in case of conditional ac	lmission to the M	. Sc. study pi	rogramme Data
	Science		1		
9	Respons	-	Department		
	Prof. Dr.	K. Ickstadt, Prof. Dr. C. Jentsch	Statistics		

Module MD Req5: Inference							
Rota	Duration	Semester	ECTS points	Workload			
every semester	1 semester	beginning of	5	150			
		programme					

1	Module	structure							
Ì	No.	Lecture/Course	Туре	ECTS	Hours				
	1a	Reading Course Inference	reading	5					
			course						
	1b	First half of "Schätzen und Testen" (	see L+T	5	2 + 1				
		BD 9, Bachelor Data Science; in							
		German)							
2									
	English (1	La) or German (1b)							
3	Content								
	One can	choose to attend either the English la	nguage reading co	ourse or the	German				
	language	lecture. Both cover the same content	s:						
	• Pa	arametric point estimation: method o	f moments and m	aximum lik	elihood;				
	СС	onsistency; sufficiency; error, bias and	loss; completene	ss; Rao-Cra	mer-bound;				
	in	variance; Bayesian estimation.							
	• Pa	arametric interval estimation: confide	nce intervals, esp	ecially for N	Iormal				
	di	istribution parameters, finding metho	ds, Bayesian estin	nation.					
	• Te	ests of hypotheses: simple and compo	site hypotheses, l	oss functio	n, (uniformly)				
	m	lost powerful tests, unbiased tests, te	sts for (multivaria	te) Normal	distribution				
	parameters, Chi-square tests, relation to confidence intervals. Literature: Alexander M. Mood, Franklin A. Graybill, Duane C. Boes: Introduction to the Theory of								
	Statistics	. McGraw-Hill 1974: Chapters VII, VIII,	IX.1-IX.6.						
4	Compete	nces / Qualification Goals							
	Students	calculate point and interval estimator	rs and carry out si	gnificance t	ests. They prove				
	basic pro	perties of estimators and tests.							
	Students	apply the methods to real data.							
5	Examinat								
		exam. A combined exam on the modu	les MD Req5 Infer	ence and N	1D Req6 Linear				
	Models is	s possible.							
6	Formality	y of examination							
	Examinat	ion based on the book chapters or or	n the lecture, resp	ectively.					
7	Module	requirements (prerequisites)							
	- none -								
8	Allocatio	n to curriculum							
8		<b>n to curriculum</b> requirement in case of conditional ad	mission to the M.	Sc. study pi	rogramme Data				
8			mission to the M.	Sc. study pi	rogramme Data				
8 9	Possible	requirement in case of conditional ad	mission to the M. Department	Sc. study pi	rogramme Data				
	Possible Science <b>Responsi</b>	requirement in case of conditional ad		Sc. study pi	rogramme Data				

Ro	ta		Duration	Semester		ECTS p	oints	Workload	
every semester			1 semester	beginning of programme		5		150	
L	Module	structure	2						
	No.	Lecture	e/Course		Туре		ECTS	Hours	
	1a	Reading	g Course Linear N	Aodels	readir course	-	5		
	1b		half of "Schätzei 9, Bachelor Data n)		L+T		5	2 + 1	
2	<b>Languag</b> English	-	rman (1b)						
1	Literatu     Thomas     and App     Compet     Student     context     Student     Student     Student	ntroducti models, bi inear mo Parameter Hypothesi ntervals. Model cho re: Kneib, St blications. s calculato of the line s apply th ation exam. A o	Springer 2015: C Qualification Goa e point and inter- ear model. They e methods to rea combined exam o	models: real dat odels. parameters, co fficients and err dence intervals: ection, predictio g Fahrmeir, Bria hapters 1, 2.1-2 Is val estimators a have knowledge al data.	variates, or varian F-Tests, n evalua n D. Mar .3, 3. nd carry on mod	residual ice. confider tion, crit rx: Regre out sign el select	ls, assur nce regi teria. ession: N ificance ion.	mptions. ons, prediction Models, Methods tests in the	
5	Models is possible.  Formality of examination Examination based on the book chapters or on the lecture, respectively.  Module requirements (prerequisites)								
3		on to curi		nditional admis	sion to th	ne M. Sc	. study	programme Data	
)	Respon	-	Prof. Dr. K. Icksta		partmer	nt			

	odule MD	Req7: Intr	roductory Case Stu	dies							
Ro eve	<b>ta</b> ery semes	ster	Duration 1 semester	Semester beginning of programme		<b>ECTS</b> 5	points	<b>W</b> 15	<b>orkload</b> 0		
1	Madula	structure									
T	No.		/Course		Туре		ECTS		Hours		
	1a		ctory Case Studies		Р		5		2		
	10 1b		"Fallstudien I" (see		P		5		4		
	10		or Data Science, 3 o						- (for 1/2 of		
			are required)						the sem.)		
2	Languag		<u></u>		I		l				
		-	rman (1b)								
3	Content		(								
	On can	choose to	either attend the	English course I	ntroduc	tory Ca	ase Studi	es o	r take part in		
			e "Fallstudien I". I	-		•			•		
		have to b			, ,			`			
	• •		urse is to familiaris	e students wit	h the ind	depend	ent eval	uatio	on of		
			ts. In addition to th			•					
			n, a central learnir	•		-			•		
			pproach and the ev			•	•				
			rning goals, studen								
	on three	e projects.	The time frame fo	or each project	is three	to six v	veeks, de	eper	nding on the		
	level of	difficulty.	The intermediate	and final result	s of the	statisti	cal evalu	atio	n are		
	present	ed alterna	tely by the groups	. After complet	ion of e	ach pro	oject, ead	ch st	udent must		
	submit a	a short, wi	ritten report in wh	ich the results	achieve	d in the	e group a	nd t	he		
	method	ology used	d are presented in	an appropriate	e manne	r.					
4	Compet	ences / Q	ualification Goals								
	Student	s work ind	lependently accord	ding to scientifi	c criteri	a and r	eport or	ally a	and in writing		
	on their	work. Stu	idents apply statist	tical methods to	o real da	ata sets	, modify	the	methods if		
		-	on their work. Students apply statistical methods to real data sets, modify the methods necessary and work out methods unknown to them. They derive solutions to problems								
		on them. T			-	reflect on them. They work together in groups. They prepare and give presentations,					
	-	explaining statistical methods and communicating results. They discuss their own and o							ations,		
	methods, results and reports with others. They complete the projects within a sho						give pres	eir o	ations, wn and other		
		-	hey work together cal methods and co	r in groups. The ommunicating	ey prepa results.	re and They di	give pres	eir o	ations, wn and other		
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## Module MD Req8: Introduction to Statistical Learning

Rota	Duration	Semester	ECTS points	Workload
every semester	1 semester	beginning of	10	300
		programme		

1	Module structure					
	No.	Lecture/Course	Туре	ECTS	Hours	
	1a	Reading Course Introduction to	reading	10		
		Statistical Learning	course			
	1b	Einführung in das statistische Lernen	L+T	10	4 + 2	
		(see BD 10, Bachelor Data Science; ir	า			
		German)				
2	Language					
	English (1a) or German (1b)					
3	Content					
	One can choose to attend either the English language reading course or the German					
	language lecture. Both cover the same contents:					
	The course Introduction to Statistical Learning is an introduction to statistical and machine					
	learning models. First, basic methods from the areas of data pre-processing and data					
	understanding are presented. Next, depending on the target variable (continuous vs.					
	discrete; observable vs. unobservable), four basic paradigms of methods are distinguished					
	and presented: Regression, Classification, Clustering and Outlier methods. For each of these					
	paradigms, different modelling concepts are presented, such as nearest neighbour methods,					
	linear methods, trees, as well as probabilistic, ensemble or regularization methods. In					
	<ul> <li>addition, general model validation and model selection methods are presented for the different machine learning paradigms.</li> <li>Literature: <ul> <li>James, G., Witten, D., Hastie, T., &amp; Tibshirani, R. (2013). An introduction to statistical</li> </ul> </li> </ul>					
	learning, Chapters 1, 2.1-2.2, 3.1-3.2, 4.1-4.4, 5.1-5.2, 6.2, 8.1-8.2					
	• Han, J & Kamber, M: Data Mining: Concepts and Techniques, Chapters 2.4, 3.4, 3.5,					
	6, 9.5, 10, 12					
4	Competences / Qualification Goals					
	Students understand different models and analysis procedures. In realistic data situations,					
	they can select suitable procedures and apply them with machine learning and statistical					
	learning toolkits. They understand the underlying mathematical-statistical theory.					
5	Examination Module exam.					
6	Formality of examination					
	Examination based on the book chapters or on the lecture.					
7	Module requirements (prerequisites)					
	- none -	- none -				
8		Allocation to curriculum				
	Possible requirement in case of conditional admission to the M. Sc. study programme Data					
	Science					
9	Responsi	-	Department			
	Prof. Dr.	A. Groll, Prof. Dr. E. Müller	Computer Scien	ce, Statistics	5	