

SYLLABUS

1. Information regarding the programme

1.1 Higher education institution	Babes-Bolyai University of Cluj-Napoca
1.2 Faculty	Faculty of Mathematics and Computer Science
1.3 Department	Department of Computer Science
1.4 Field of study	Computer Science
1.5 Study cycle	Master
1.6 Study programme / Qualification	Advanced Computational Intelligence

2. Information regarding the discipline

2.1 Name of the discipline	Numerical Modelling in Data Analysis / Modelare Numerică în Analiza Datelor						
2.2 Course coordinator	Prof. Dr. Lehel Csató						
2.3 Seminar coordinator	Prof. Dr. Lehel Csató						
2.4. Year of study	1	2.5 Semester	1	2.6. Type of evaluation	E	2.7 Type of discipline	Compulsory
2.8 Code	MME8172						

3. Total estimated time (hours/semester of didactic activities)

3.1 Hours per week	3	Of which: 3.2 course	2	3.3 seminar/laboratory	1
3.4 Total hours in the curriculum	42	Of which: 3.5 course	28	3.6 seminar/laboratory	14
Time allotment:					hours
Learning using manual, course support, bibliography, course notes					42
Additional documentation (in libraries, on electronic platforms, field documentation)					24
Preparation for seminars/labs, homework, papers, portfolios and essays					28
Tutoring					12
Evaluations					4
Other activities:					-
3.7 Total individual study hours	108				
3.8 Total hours per semester	150				
3.9 Number of ECTS credits	6				

4. Prerequisites (if necessary)

4.1. curriculum	Algebra and calculus, knowledge of python / matlab / julia
4.2. competencies	<ul style="list-style-type: none"> • Basic mathematics • Using computers for programming.

5. Conditions (if necessary)

5.1. for the course	<ul style="list-style-type: none"> • Using of online TEAMS application.
5.2. for the seminar /lab activities	<ul style="list-style-type: none"> • Students will use – if possible – their own laptop

6. Specific competencies acquired

Professional competencies	<ul style="list-style-type: none"> • Understanding the mathematical concepts used in data modelling. • CE1.3 - Using methods from artificial intelligence in solving real-world problems. • CE3.4 - Analysis and modelling of data.
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Transversal competencies	<ul style="list-style-type: none"> • CT1. - The ability to apply intelligent data analysis methods in solving real world problems. • CT3 - The usage of efficient methods and techniques that facilitate the learning, the information, the research, and the development process.
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7. Objectives of the discipline (outcome of the acquired competencies)

7.1 General objective of the discipline	Familiarization with the mathematics and numerical methods that can be used in machine learning.
7.2 Specific objective of the discipline	<ul style="list-style-type: none"> - to highlight the use of the numerical methods in data analysis - to familiarize with programming languages that implement these methods, - to highlight the need for understanding the mathematics behind data analysis methods.

8. Content

8.1 Course	Teaching methods	Remarks
<ul style="list-style-type: none"> • Week 1: Administration and organization Introducing the objectives of the course. 	<ul style="list-style-type: none"> • Interactive exposure • Explanation 	
<ul style="list-style-type: none"> • Week 2: Mathematical background: the convolution, the notions related to probabilities. 		
<ul style="list-style-type: none"> • Week 3: Maximum Likelihood and the respective geometric interpretations. Illustrations on toy and real data. The MAP method 		
<ul style="list-style-type: none"> • Week 4: Bayesian parameter estimation 		
<ul style="list-style-type: none"> • Week 5: The classification problem and different likelihood functions 		
<ul style="list-style-type: none"> • Week 6: Approximating the likelihood functions. The first and second order methods. Variational methods. 		
<ul style="list-style-type: none"> • Week 7: Unsupervised methods and the manifold hypothesis. 		
<ul style="list-style-type: none"> • Week 8: The Principal Component Analysis 		
<ul style="list-style-type: none"> • Week 9: Probabilistic Principal Components. Comparisons and applications. 		
<ul style="list-style-type: none"> • Week 10: Independent Component Analysis. Applications 		
<ul style="list-style-type: none"> • Week 11. Clustering methods and the EM algorithm 		

<ul style="list-style-type: none"> • Week 12, 13, 14. Presentations related to other advanced methods: <ul style="list-style-type: none"> • SVN • Gaussian Processes, • Ensemble methods 		
Bibliography [1] Bishop C.M (2006) Pattern Recognition and Machine Learning, Springer Verlag freely available at: https://www.microsoft.com/en-us/research/people/cmbishop/prml-book/ [2] Deisenroth M.P, Faisal A.A, Soon Ong C (2020) Mathematics for Deep learning, Cambridge University Press freely available at: https://mml-book.github.io/ [3] Li M, Lipton Z.C, Smola A.J, Zhang A (2020): Dive into Deep Learning, Online book, release 0.14.4, freely available at: https://d2l.ai/ accessed 04.01.2020		

8.2 Seminar / laboratory	Teaching methods	Remarks
1.Administration. presenting the working environment, setting up the topics for presentation	<ul style="list-style-type: none"> • Interactive exposure • Explanation • Conversation 	
2-3. Tutoring related to the presentation topics	<ul style="list-style-type: none"> • Interactive exposure • Explanation • Conversation 	
4-5. Tutoring and scoring the progress of the presentations	<ul style="list-style-type: none"> • Interactive exposure • Explanation • Conversation 	
6-7. Final presentations	<ul style="list-style-type: none"> • Interactive exposure • Explanation • Conversation 	
Bibliography [4] Hastie T, Tibshirani R, Friedman R (2009) The Elements of Statistical Learning, Springer Verlag freely available at: https://web.stanford.edu/~hastie/ElemStatLearn/ [5] Haykin S (2009) Neural Networks and Learning Machines, Third Edition, Pearson Education. [6] Murphy K.M (2012) Machine Learning, a Probabilistic Perspective, The MIT Press.		

9. Corroborating the content of the discipline with the expectations of the epistemic community, professional associations and representative employers within the field of the program

The content of the discipline is consistent with the similar disciplines from international universities from abroad – e.g. Stanford, ELTE. It also confirms to the requirements from potential employers asking for intelligent data analysts.

10. Evaluation

Type of activity	10.1 Evaluation criteria	10.2 Evaluation methods	10.3 Share in the grade
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10.4 Course	Correctness of the accumulated knowledge.	Written exam (in the regular session)	40%
10.5 Seminar/lab activities	LAB activity	Grade awarded pro rata	10%
	Laboratory exercises	Evaluation of the work that was handed in during the semester	50%
10.6 Minimum performance standards			
All three – COURSE, SEMINAR, ASSIGNMENTS have to be at least 60%.			
The grading (the conversion of percentages to final note):			
<ul style="list-style-type: none"> • 60-66 – 5 • 67-73 – 6 • 74-80 – 7 • 81-87 – 8 • 88-94 – 9 • 95-100 – 10 			

Date
27.04.2020

Signature of course coordinator
Prof. dr. Lehel CSATÓ

Signature of seminar coordinator
Prof. dr. Lehel CSATÓ

Date of approval

Signature of the head of department
Prof. dr. Anca Andreica

Note that the unforeseen situation due to the pandemic requires that all exams will be taken ONLINE, the presence at the seminars/LAB classes – in 90% - is compulsory, and that the final examination will be

1. The course examination will be on TEAMS and will have two stages:
 - a. Preparation of a NOTEBOOK (in Julia language) that solves a problem,
 - b. “Oral” examination in TEAMS – of about 30 mins – discussing about the concepts taught during the classes.
2. The lab examination will be through a project and periodic assessments – online.