#### **SYLLABUS**

### 1. Information regarding the programme

1.1 Higher education institution	Babeş Bolyai University
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	Bachelor
1.6 Study programme / Qualification	Artificial Intelligence

# 2. Information regarding the discipline

2.1 Name of the discipline (en)			Knowledge based systems				
(ro)							
2.2 Course coordinator			Lect. Univ. Dr. Găceanu Radu				
2.3 Seminar coordinator		Lect. Univ. Dr. Găceanu Radu					
2.4. Year of study	3	2.5 Semester	5	2.6. Type of	E	2.7 Type of	Compulsory
				evaluation		discipline	
2.8 Code of the		MLE5201					
discipline							

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

<u> </u>					
3.1 Hours per week	3	Of which: 3.2 course	2	3.3	1
				seminar/laboratory	
3.4 Total hours in the curriculum	42	Of which: 3.5 course	28	3.6	14
				seminar/laboratory	
Time allotment:					
Learning using manual, course support, bibliography, course notes					
Additional documentation (in libraries, on electronic platforms, field documentation)					
Preparation for seminars/labs, homework, papers, portfolios and essays					
Tutorship					6
Evaluations					2
Other activities:					-
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3.7 Total individual study hours	58
3.8 Total hours per semester	100
3.9 Number of ECTS credits	4

# **4. Prerequisites** (if necessary)

4.1. curriculum	Algorithms, data structures, statistics
4.2. competencies	Average programming skills

# **5. Conditions** (if necessary)

5.1. for the course	• Projector
5.2. for the seminar /lab	Computers, specific development environment
activities	

6. Specific competencies acquired

	te competencies acquired
Professional competencies	CE1.1 Description of artificial intelligence concepts and research directions  CE1.2 Evaluation of the quality and stability of the obtained solutions and their comparison with the solutions obtained by traditional methods  CE1.3 Using artificial intelligence methods, techniques and algorithms to model solutions to classes of problems
Transversal competencies	CT1. Application of efficient work rules and responsible attitudes towards the scientific domain, for the creative exploitation of one's own potential according to the principles and rules of professional ethics  CT3. Use of efficient methods and techniques for learning, information, research and development of abilities for knowledge exploitation, for adapting to the needs of a dynamic society and for communication in a widely used foreign language.

# 7. Objectives of the discipline (outcome of the acquired competencies)

7.1 General objective of the discipline	To introduce the student in Knowledge-based systems (KBS)
7.2 Specific objective of the discipline	This course is aimed to advance both theoretical and practical aspects of KBS. The course aims to provide an overview of the discipline and its main areas. At the end of the course, students will understand the basic principles of KBS and associated algorithmic approaches and have knowledge of KBS applications.

### 8. Content

8.1 Course	Teaching methods	Remarks
1. Introduction to KBS	<ul> <li>Interactive</li> </ul>	
2. KBS	exposure	
a. Components	<ul> <li>Presentation</li> </ul>	
b. Knowledge representation	<ul> <li>Explanation</li> </ul>	

<ul> <li>Formal logic</li> <li>Rules</li> <li>Semantic networks</li> <li>c. Inference process</li> <li>Techniques – certainty</li> <li>a. Logic</li> <li>b. Rule-based</li> <li>Techniques – uncertainty</li> <li>a. Probabilities</li> <li>b. Fuzzy</li> <li>3. Logic-based KBS</li> <li>4. KBS – certainty</li> <li>a. Design</li> <li>b. Architecture</li> <li>Knowledge base</li> <li>Inference</li> <li>a. Forward</li> <li>b. Backward</li> <li>c. Conflicts</li> <li>5. KBS – uncertainty</li> <li>a. Architecture</li> <li>Knowledge base</li> <li>Inference</li> <li>a. Forward</li> <li>b. Backward</li> <li>c. Conflicts</li> <li>5. KBS – uncertainty</li> <li>a. Architecture</li> <li>Knowledge base</li> <li>Inference</li> <li>a. Bayes</li> <li>b. Certainty theory</li> <li>c. Fuzzy logic – stages</li> <li>o. Fuzzification</li> <li>o. Rules</li> <li>o. Fuzzy inference</li> <li>o. Aggregation</li> <li>o. Defuzzification</li> <li>o. Results</li> <li>6. Strengths and weaknesses of KBS</li> <li>7. Real-world KBSs</li> <li>Bibliography</li> <li>I. S. Russell. P. Norvig. Artificial Intelligence: A Modern Approach Prentice Hall, 1995</li> </ul>	T 11 '		
Semantic networks  c. Inference process  Techniques – certainty a. Logic b. Rule-based Techniques – uncertainty a. Probabilities b. Fuzzy  3. Logic-based KBS 4. KBS – certainty a. Design b. Architecture  Knowledge base Inference a. Forward b. Backward c. Conflicts  5. KBS – uncertainty a. Architecture  Knowledge base Inference a. Forward b. Backward c. Conflicts  5. KBS – uncertainty a. Architecture  Knowledge base Finference a. Bayes b. Certainty theory c. Fuzzy logic – stages Fuzzification o Rules o Fuzzy inference o Aggregation o Defuzzification o Results  6. Strengths and weaknesses of KBS 7. Real-world KBSs  Bibliography   Case-study discussions			
c. Inference process  Techniques – certainty a. Logic b. Rule-based Techniques – uncertainty a. Probabilities b. Fuzzy  3. Logic-based KBS 4. KBS – certainty a. Design b. Architecture  Knowledge base Inference a. Forward b. Backward c. Conflicts  5. KBS – uncertainty a. Architecture  Knowledge base Inference a. Forward b. Backward c. Conflicts  5. KBS – uncertainty a. Architecture  Fuzzy logic – stages b. Certainty theory c. Fuzzy logic – stages  Fuzzification Rules Fuzzy inference Aggregation Defuzzification Results  6. Strengths and weaknesses of KBS 7. Real-world KBSs  Bibliography		examples	
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- 2. C. Groşan, A. Abraham, Intelligent Systems: A Modern Approach, Springer, 2011
- 3. A. Hopgood, Intelligent Systems for Engineers and Scientists, CRC Press, 2001
- 4. H.F. Pop, G. Şerban, Inteligență artificială, Cluj Napoca, 2004
- 5. D. J. C. MacKey, Information Theory, Inference and Learning Algorithms, Cambridge University Press, 2003
- 6. G.J. Klir, B. Yuan, Fuzzy Sets and Fuzzy Logic, Prentice Hall, 1995

8.2 Seminar / laboratory	Teaching methods	Remarks
L1: Eficient solutions for algorithmic problems	<ul> <li>Interactive</li> </ul>	
L2-L3: Design and implementation of KBS – certainty	exposure	
L4-L5: Design and implementation of KBS – uncertainty	<ul> <li>Explanation</li> </ul>	
L6-L7: Recommendation systems	<ul> <li>Conversation</li> </ul>	
	<ul> <li>Didactical</li> </ul>	
	demonstration	

#### Bibliography

- 1. S. Russell, P. Norvig, Artificial Intelligence: A Modern Approach, Prentice Hall, 1995
- 2. C. Groşan, A. Abraham, Intelligent Systems: A Modern Approach, Springer, 2011
- 3. A. Hopgood, Intelligent Systems for Engineers and Scientists, CRC Press, 2001
- 4. G.J. Klir, B. Yuan, Fuzzy Sets and Fuzzy Logic, Prentice Hall, 1995

# 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 course exists in the curriculum of many universities in the world.
- The results of course are considered by software companies particularly useful and topical, developing needed abilities in modelling and visualization of data.

#### 10. Evaluation

Type of activity	10.1 Evaluation criteria	10.2 Evaluation methods	10.3 Share in the grade (%)
10.1 Course	Knowledge of the basic concepts of the field Applying the intelligent principles from the course content to solve complex and difficult problems	Final written report	10%
10.2 Seminar/lab activities	· Specification, design, implementation and testing of intelligent methods; problem solving; written reports and presentations	Observation of the student while solving the task Practical projects Written reports and presentations during the semester	90%

#### 10.3 Minimum performance standards

- Each student has to demonstrate that he has reached an acceptable level of knowledge and understanding of the field, that he is able to express the knowledge in a coherent form, that he has the ability to establish certain connections and to use the knowledge in solving some problems.
- To pass the exam you must:
- at least 60% of the assignments during the semester are completed
- an evaluation average (written exam, seminar, laboratory) to be above 5

Date	Signature of course coordinator	Signature of seminar coordinator
	Lect. PhD. Găceanu Radu	Lect. PhD. Găceanu Radu

 Assoc. Prof. PhD. Sterca Adrian