#### **SYLLABUS**

in mormation regarding the programme			
Babes-Bolyai University			
Mathematics and Computer Science			
Computer Science			
Computer Science			
Undergraduate (3 <sup>rd</sup> year bachelor)			
	Babes-Bolyai University         Mathematics and Computer Science         Computer Science         Computer Science		

#### 1. Information regarding the programme

## 2. Information regarding the discipline

2.1 Name of the di	the discipline (en) Introduction to Big Data						
(ro)	ro) Introducere in Big Data						
2.2 Course coordin	rse coordinator Lect. dr. Ioana Ciuciu						
2.3 Seminar coordinator		Lect. dr. Ioana Ciuciu					
2.4. Year of study	3	2.5 Semester	6	2.6. Type of evaluation	C	2.7 Type of discipline	optional
2.8 Code of the discipline		MLE5093					

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

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:	•		•		hours
Learning using manual, course support	rt, bił	oliography, course note	S		36
Additional documentation (in libraries, on electronic platforms, field documentation)					36
Preparation for seminars/labs, homework, papers, portfolios and essays					36
Tutorship					5
Evaluations					14
Other activities:					
3.7 Total individual study hours 127					
3.8 Total hours per semester		175			
3.9 Number of ECTS credits		4			

## 4. Prerequisites (if necessary)

4.1. curriculum	-
4.2. competencies	- programming skills

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

5.1. for the course	
5.2. for the seminar /lab	Laboratory with computers; Hadoop, SQL Server, SQL Server
activities	Management Studio

#### 6. Specific competencies acquired

	competencies acquirea
ional encies	• Use of non-traditional databases for storing and processing large amounts of data
Professional competencie	Advanced querying over distributed information resources
Profess compet	• Evaluation, testing and validation with real-world data
	• Methods and algorithms for data processing and analysis applied to Big Data
es	• Multidisciplinary competencies spanning various application sectors (e.g., life sciences and bioinformatics, telco, media, finance, security, health, energy, etc.)
Transversal competencies	• Data Science competencies, combining data analyst and data specialist- specific competencies (e.g., competencies from the fields of mathematics, statistics, information science, computer science, databases, machine learning, data mining, visualization, etc.)

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

7.1 General objective of the discipline	• Handling (extremely) large amounts of digital data in various formats (text, video, financial, medical, etc.)	•
7.2 Specific objective of the discipline	<ul> <li>Enable the use of novel algorithms, software infrastructures and methodologies for the purpose of handling (store, retrieve, analyze) large amounts of data</li> <li>Provide decision support over large volumes of data</li> <li>Enable the creation of applications and services for various business domains based on the results of big data analysis.</li> </ul>	•

8. Content		
8.1 Course	Teaching methods	Remarks
1. Introduction to Data Science and Big Data	Exposure, description,	
	explanation, examples,	
	case studies	
2-3. Big Data Architecture	Exposure, description,	
	explanation, examples,	
	case studies	
4. Data Storage	Exposure, description,	
	explanation, examples,	
	case studies	

5. Industrial Standards for Data Mining Projects	Exposure, description,
	explanation, examples,
	case studies
6-7. Data Integration	Exposure, description,
	explanation, examples,
	case studies
8-9. Data Warehousing	Exposure, description,
	explanation, examples,
	case studies
10. Data Visualization	Exposure, description,
	explanation, examples,
	case studies
11. Big Data Project Proposals Presentation	Exposure, description,
	explanation, examples,
	case studies
12. Big Data Case Studies	Discussions around case
	studies
13-14. Student Presentations	Exposure, description,
	explanation, examples,
	case studies
Rihliogranhy	

#### Bibliography

Frontiers in Big Data Analysis, The National Academies Press, Washington, prepublication draft

V. Agneeswaran, Big Data Analytics Beyond Hadoop, Pearson Education, 2014

T. White, Hadoop: The Definitive Guide, O'Reilly, 2009

D. Miner, A. Shook, MapReduce Design Patterns: Building Effective Algorithms and Analytics for Hadoop and Other Systems, O'Reilly, 2012

P. K. Janert, Data Analysis with Open Source Tools, O'Reilly, 2010

*Q. E. McCallum, Bad Data Handbook: Cleaning Up The Data So You Can Get Back To Work, O'Reilly, 2012* 

O'Reilly Radar Team, Big Data Now: Current Perspectives from O'Reilly Radar, 2011

S.T. Allen, Storm Applied, 2015

M. Hamstra, Learning Spark, 2014

M. Barlow, Real-Time Big Data Analytics: Emerging Architecture, O'Reilly Media, 2013

J. Janssens, Data Science at the Command Line: Facing the Future with Time-Tested Tools, O'Reilly, 2014

T. Ojeda et al., Practical Data Science Cookbook, 2014

Data Science and Big Data Analytics, EMC Education Services, 2014

R. Morisson, Big Data Now, 2014

G. De Francisci Morales, Big Data and the Web: Algorithms for Data Intensive Scalable Computing IMT Institute for Advanced Studies, 2012

K Asanivik et al., The Landscape of Parallel Computing Research: A View from Berkeley, 2006

J. Dean, Big Data, Data Mining and Machine Learning: Value Creation for Business Leaders and Practitioners, Wiley, 2014

R. Glass and s. Callahan, The Big Data-Driven Business: How to Use Big Data to Win Customers, Beat Competitors, and Boost Profits, Wiley, 2014

D.L. Herben, Big Data, Big Analytics: Emerging Business Intelligence, 2014

A. M. Paganoni and P. Secchi, Advances in Complex Data Modeling and Computational Methods in Statistics, Springer, 2014

8.2 Seminar / laboratory	Teaching methods	Remarks
Semester project organized with groups of 3-4	Team work	Project groups will be
students	Individual work	monitored via a project

	Periodic meetings with the lab responsible Periodic deliverables	wiki managed by the course/lab responsible		
Bibliography				
http://mahout.apache.org/				
http://www.tutorialspoint.com/mahout/mahout_introduc	tion.htm			
http://spark.apache.org/documentation.html				
http://shark.cs.berkeley.edu/				
http://spark.apache.org/				
http://nosql-database.org/				
https://www.mongodb.com/nosql-explained				

# **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

Synergies with various local and EU innitiatives: local industry, European Data Science Academy (EDSA), EU projects such as FERARI, LIFT, LOD2, Open Data Monitor, Data Publishing through the Cloud, Trendminder, Web Observatory, etc.

#### **10. Evaluation**

Type of activity	10.1 Evaluation criteria	10.2 Evaluation methods	10.3 Share in the grade (%)		
10.4 Course	<ul> <li>to be familiar with the main concepts of the domain</li> <li>to be able to apply these principles in real-life use cases</li> </ul>	Written exam	50%		
10.5 Seminar/lab activities	<ul> <li>Big Data processing</li> <li>Big Data storage</li> <li>Big Data integration</li> <li>Applications</li> </ul>	Laboratory work	50%		
10.6 Minimum performance standards					
A minimum grade of 5 is necessary for both the written exam and the practical work					

Date

Signature of course coordinator

Signature of seminar coordinator

05.05.2017

Lect. Dr. Ioana Ciuciu

Lect. Dr. Ioana Ciuciu

Date of approval

Signature of the head of department

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