#### **SYLLABUS**

1.1 Higher education	Babes-Bolyai University			
institution				
1.2 Faculty	Mathematics and Computer Science			
1.3 Department	Computer Science			
1.4 Field of study	Computer Science			
1.5 Study cycle	Undergraduate (3 <sup>rd</sup> year bachelor)			
1.6 Study programme /				
Qualification				

#### 1. Information regarding the programme

## 2. Information regarding the discipline

2.1 Name of the disciplineIntroduction to Big Data Analysis							
2.2 Course coordinator Ciuciu Ioana							
2.3 Seminar coordinator				Ciuciu Ioana			
2.4. Year of	3	2.5	5	2.6. Type of	Ε	2.7 Type of	optional
study		Semester		evaluation		discipline	

### 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, bibliography, course notes					
Additional documentation (in libraries, on electronic platforms, field documentation)					
Preparation for seminars/labs, homework, papers, portfolios and essays					
Tutorship					
Evaluations					
Other activities:					-
3.7 Total individual study hours					•
3.8 Total hours per semester					

#### 4. Prerequisites (if necessary)

3.9 Number of ECTS credits

4.1. curriculum	
4.2. competencies	

4

## 5. Conditions (if necessary)

5.1. for the course	
5.2. for the seminar /lab	
activities	

## 6. Specific competencies acquired

SS I	• Use of non-traditional databases for storing and processing large amounts of data
<b>Professional</b> competencies	<ul> <li>Advanced querying over distributed information resources</li> <li>Evaluation, testing and validation with real-world data</li> </ul>
	• Methods and algorithms for data processing and analysis applied to Big Data
l 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</li> </ul>	
	domains based on the results of big data analysis.	

## 8. Content

8. Content					
8.1 Course	Teaching methods	Remarks			
<ul> <li>Introduction to Data Science and Big Data</li> </ul>	<ul> <li>Interactive exposure</li> <li>Explanation</li> <li>Conversation</li> <li>Didactical demonstration</li> </ul>				
Programming tools for data science and data visualization	<ul> <li>Interactive exposure</li> <li>Explanation</li> <li>Conversation</li> <li>Didactical demonstration</li> </ul>				
Basics of NoSQL data management solutions	<ul> <li>Interactive exposure</li> <li>Explanation</li> <li>Conversation</li> <li>Didactical demonstration</li> </ul>				
Basic analytics	<ul> <li>Interactive exposure</li> <li>Explanation</li> <li>Conversation</li> <li>Didactical demonstration</li> </ul>				
Basic Machine Learning	Interactive exposure				

	• Explanation
	Conversation
	Didactical
	demonstration
Basic Data Mining	• Interactive exposure
	• Explanation
	Conversation
	Didactical
	demonstration
Big Data Architecture	Interactive exposure
	• Explanation
	Conversation
	Didactical     demonstration
Dia Data Analyzia	
Big Data Analysis	• Interactive exposure
	• Explanation
	Conversation
	Didactical
	demonstration
Overview of Apache Mahout, Spark, Storm	• Interactive exposure
& Shark	• Explanation
	Conversation
	• Didactical
	demonstration
Information Systems concepts	Interactive exposure
F	• Explanation
	Conversation
	Didactical
	demonstration
Eundemontale of Data Viewalization	
• Fundamentals of Data Visualization	Interactive exposure
	• Explanation
	Conversation
	• Didactical
	demonstration
• Introduction to Semantics and Linked Data	• Interactive exposure
for Data Science	• Explanation
	Conversation
	Didactical
	demonstration
Bibliography	

#### **Bibliography**

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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
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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 students	<ul> <li>Team work</li> <li>Individual work</li> <li>Periodic meetings with the lab responsible</li> <li>Periodic deliverables</li> <li>Project groups will be monitored via a project wiki managed by the course/lab responsible</li> </ul>				
Bibliography					
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http://spark.apache.org/documentation.html	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**

Project-based evaluation	• Project groups will present and demonstrate their semester project	
	•	
Written exam	<ul> <li>Evaluation of he theoretical aspects</li> <li>Evaluation of the targeted competencies</li> </ul>	

	via exercises	
Date	Signature of course coordinator	Signature of seminar coordinator
20 Nov. 2015	I. Ciuciu	I. Ciuciu
D	Date of approval	Signature of the head of department