SYLLABUS

1. Information regarding the programme

1.1 Higher education institution	**Babeş-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	**High Performance Computing and Big Data Analytics**

2. Information regarding the discipline

2.1 Name of the discipline (en) (ro)			**Big Data Processing and Applications**				
2.2 Course coordinator			**Lect. Dr. Ioana-Georgiana Ciuciu**				
2.3 Seminar coordinator			**Lect. Dr. Ioana-Georgiana Ciuciu**				
2.4. Year of study	**2**	2.5 Semester	**3**	2.6. Type of evaluation	**E**	2.7 Type of discipline	**Compulsory**
2.8 Code of the discipline		MME8158					

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

3.1 Hours per week	4	Of which: 3.2 course	2	3.3 seminar/laboratory	2
3.4 Total hours in the curriculum	56	Of which: 3.5 course	28	3.6 seminar/laboratory	28
Time allotment:					hours
Learning using manual, course support, bibliography, course notes					35
Additional documentation (in libraries, on electronic platforms, field documentation)					45
Preparation for seminars/labs, homework, papers, portfolios and essays					47
Tutorship					15
Evaluations					16
Other activities:					-

3.7 Total individual study hours	119
3.8 Total hours per semester	175
3.9 Number of ECTS credits	8

4. Prerequisites (if necessary)

4.1. curriculum	
4.2. competencies	• Basic knowledge of data analytics, preferably

	• Basic knowledge of data visualization, preferably • Programming skills

5. Conditions (if necessary)

5.1. for the course	• Room with video projector
5.2. for the seminar /lab activities	• Room with computers as needed; • Big Data software installed • High level programming language environment

6. Specific competencies acquired

Professional competencies	• Use of non-traditional databases for storing and processing large amounts of data • Advanced querying over distributed information resources • Evaluation, testing and validation with real-world data • Learning to conduct incipient research in the field of Big Data
Transversal competencies	• Methods and algorithms for data processing and analysis applied to Big Data • Multidisciplinary competencies spanning various application sectors (e.g., life sciences and bioinformatics, telco, media, finance, security, health, energy, etc.) • Data Science competencies, combining data analyst and data engineer- specific competencies (e.g., competencies from the fields of mathematics, statistics, information science, computer science, databases, machine learning, data mining, visualization, etc.) • Manifest responsible attitudes towards the scientific and didactic fields

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	• Enable the use of novel algorithms, software infrastructures and methodologies for the purpose of processing (store, retrieve, analyze) large amounts of data • Provide decision support over large volumes of data • Enable the creation of applications and services for various business domains based on the results of big data analysis.

8. Content

8.1 Course	Teaching methods	Remarks
1. Introduction to Data Science and Big Data	Exposure, description, explanation, examples,	Data Science main concepts, the Data Science Process,

	case studies	challenges, data availability, data types, tools
2. Industrial Standards for Data Mining Projects	Exposure, description, explanation, examples, case studies	Methodology for Data Science projects (CRISP-DM)
3. Big Data Architecture	Exposure, description, explanation, examples, case studies	Traditional database systems versus Big Data systems, the Lambda Architecture, a model for building a Big Data system, case studies and examples
4. Batch Layer	Exposure, description, explanation, examples, case studies	Big Data storage, data model for Big Data, batch computing, the Hadoop Ecosystem
5. Serving Layer - part I	Exposure, description, explanation, examples, case studies	Requirements, performance metrics, the normalization/denormalization problem, tools
6. Serving Layer - part II	Exposure, description, explanation, examples, case studies	
7. Speed Layer - part I	Exposure, description, explanation, examples, case studies	Computing and storing of real time views, real time updates, tools
8. Speed Layer - part II	Exposure, description, explanation, examples, case studies	
9. Data Ingestion	Exposure, description, explanation, examples, case studies	Definitions and design considerations, batch ingestion, real time ingestion, tools
10. NoSQL Solutions for Big Data	Exposure, description, explanation, examples, case studies	NoSQL databases, NoSQL Data Models Tutorial provided
11. Data Visualization	Exposure, description, explanation, examples, case studies	Scientific data visualization principles, visual analytics for exploratory data analysis
12. Big Data Case Studies	Exposure, description, explanation, examples, case studies	Presentation of Big Data (industrial) case studies
13. Big Data Research Essays Presentation	Exposure, description, explanation, examples, case studies	Student essay presentation
14. Big Data Research Essays Presentation	Exposure, description, explanation, examples, case studies	Student essay presentation

Bibliography

Marz, N., & Warren, J. (2015). *Big Data. Principles and Best Practices of scalable real-time systems*. Manning Publications

Cielen, D., Meysman, A.D.B., & Ali, M. (2016). *Introducing Data Science. Big Data, machine learning, and more, using Python tools*. Manning Publications

Grus, J. (2019). *Data Science from Scratch: First Principles with Python*. O'Reilly Media, Inc.

Damji, J.S., Wenig, B., Das, T., & Lee, D. (2020). *Learning Spark*. O'Reilly Media, Inc.

Sadalage, P., Fowler, M. (2013). *NoSQL Distilled: A Brief Guide to the Emerging World of Polyglot Persistence*. Pearson Education, Inc.

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

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

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

8.2 Seminar / laboratory	Teaching methods	Remarks
Semester project organized with groups of about 2-3 students (depending on the requirements and the equipment needed) Team work will be autonomous (focus on creativity and critical thinking) Technical tutorials will be provided to support student work around the most important aspects of Big Data storage and processing (e.g., Hadoop shell, PySpark, Data Ingestion with Apache Sqoop, NoSQL, etc.)	Research-informed Learning Tutorial-based Problem-solving approach Team work Big Data solutions for concrete problems and case studies	Groups will be monitored via a project wiki managed with the course/lab the responsible The lab takes place every two weeks and takes two hours
Bibliography Same as for the course		

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 initiatives: local industry, European Data Science Academy (EDSA, https://edsa-project.eu/),EU projects such as LETHE (https://cordis.europa.eu/project/id/101017405), FARE (https://cordis.europa.eu/project/id/853566), the Human Brain Project (https://www.humanbrainproject.eu/en/), SoBigData (http://project.sobigdata.eu/), etc.

Collaboration with the IT industry: invited lectures with real-life use cases, semester project topics, equipment (e.g., smart sensors).

10. Evaluation

Type of activity	10.1 Evaluation criteria	10.2 Evaluation methods	10.3 Share in the grade (%)
10.4 Course	- to be familiar with the main concepts of the domain -to be able to model a problem from a specific application field relying on emergent Big Data technologies - to be able to apply these principles in real-life use cases	Written exam/ Evaluation of a research essay	50%
10.5 Seminar/lab activities	- to be able to propose viable creative solutions to real-life big data challenges from various application domains - to be able to consume (query, analyze)Big Data in order to derive information relevant to use cases from various application domains - to demonstrate critical thinking - to successfully perform individual and team-based tasks	Semester project	50%
10.6 Minimum performance standards			
➢ A minimum grade of 5 (on a scale from 1 to 10) is necessary for the written exam, the practical work and the research essay ➢ The lab attendance is compulsory at a rate of 90%, according to the decision of the Computer Science Department Council (http://www.cs.ubbcluj.ro/wp-content/uploads/Hotarare-CDI-15.03.2017.pdf)			

Date	Signature of course coordinator	Signature of seminar coordinator
...30 April 2022.....	Lect. Dr. Ioana-Georgiana Ciuciu	Lect. Dr. Ioana-Georgiana Ciuciu

Date of approval

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Signature of the head of department

Prof. Dr. Laura Diosan