

MD-2068, CHISINAU, 9/7 STUDENTILOR STR, PHONE: 022 50-99-63, www.utm.md

Big Data Technologies

1. Course/Module information

Faculty	Computers, Informatics, and Microelectronics				
Chair/department	Informatics and Systems Engineering				
Study cycle	CYCLE II, Master's Studies				
Study program	Data Science				
Year of study	Semester	Evaluation type	Formative category	Optionality category	ECTS credits
I	II	E	F	O	5

2. Estimated total time

Total hours in the curriculum	Including				
	Auditory hours		Individual work		
	Lecture	Practical work	Term paper	Study of theoretical material	Application development
150	20	20	---	110	-

3. Prerequisites for access to the course/module

According to the curriculum plan	The study is based on the knowledge accumulated by students in the following subjects: probability theory, mathematical statistics, programming methods, web technologies, and databases. Additionally, master's students should be familiar with computer architecture, at least with one operating system (Windows, Linux, MacOS), as well as some programming languages such as C#, C++, HTML, PHP, and Python.
According to competencies	<p>Master's students must possess the following knowledge, competencies, and skills to master the academic discipline "Big Data Technologies": data analysis methods, application of modern Database Management Systems, including MySQL and MS SQL Server, multidimensional data modeling, modeling and application of Operational Databases and Data Warehouses, OLTP, OLAP, and BI technologies (e.g., VS, Power BI), as well as a certain level of data exploration using Data Mining tools.</p> <p>Additionally, master's students should be able to work in teams, effectively use ICT for solving applied professional problems, and apply mathematical modeling skills to socio-economic phenomena, environmental issues, technical fields, as well as domains such as finance and banking, education, healthcare, insurance, and marketing.</p> <p>The main outcomes of the "Big Data Technologies" course include applying knowledge, competencies, and skills in Data Engineering activities, as well as in the study of related disciplines such as "Data Analysis," "Data Science," "Big Data Analytics," and "Business Informatics," among others.</p>

4. Conditions for conducting the educational process

Lecture	<p>The theoretical material is presented in the lecture hall, combining the use of a projector and computer as well as a traditional whiteboard. Phone calls are not tolerated during the lecture.</p> <p>Throughout the course, students will take three generalizing quick tests on previously covered material, specifically after the 3rd, 5th, and final lecture sessions. The conditions for these tests are provided on the MOODLE platform.</p>
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Laboratory/seminar/practical	<p>For completed tasks, students must prepare reports in digital format and, if required, in hard copy (printed), following the conditions specified in the methodological guidelines. The deadline for submitting laboratory or practical work, whether individual or group-based, is determined at the lecturer's discretion.</p> <p>For late submissions, students will be penalized by 1 (one) point per week of delay.</p>
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5. Specific competencies acquired

Professional competencies	<p>CPM1: System architecture design and development</p> <p>CPM2: Monitoring technological trends. Innovation. Sustainable development</p> <p>CPM4: Staff development</p>
Transversal competencies	<p>CTM1: Autonomy and responsibility</p> <p>CTM2: Social interaction</p> <p>CPM5: Process improvement</p>

6. Course/Module objectives

General objective	<p>The acquisition of knowledge, development of skills, practical abilities, and competencies in Data Engineering/Big Data Analytics aims to facilitate their application to solving a wide range of challenges within the national economy. This is achieved through the simulation of processes and phenomena in fields relevant to future ICT specialists trained at the Faculty of Computers, Informatics, and Microelectronics, under the Data Science study program.</p> <p>Specifically, upon completing this course, master's students should be able to:</p> <ul style="list-style-type: none"> • conduct studies and evaluations on the implementation and efficiency of Big Data technologies and tools within enterprises (Business Big Data Management); • implement and apply analytical and decision-support tools based on Big Data for Decision Management; • develop new models for an enterprise's information infrastructure, considering the capabilities of Big Data technologies (Model Management).
Specific objectives	<ul style="list-style-type: none"> •to acquire and gain new knowledge regarding the use of the technologies and tools underlying the "Big Data Technologies" course, necessary for activities in related fields such as data engineering, data science, and artificial intelligence; •to familiarize with the application of emerging ICT fields, such as artificial intelligence, Big Data analytics, cloud computing, augmented and virtual reality, in innovation and deep research activities on socio-economic phenomena, in order to be prepared to apply emerging technological trends in transforming research and production models, addressing current challenges; •to develop new capabilities for analyzing and using Big Data technologies, from the perspective of data utilization efficiency, when working on a concrete informational project related to the decision-making process efficiency of an entity, where Big Data is assumed to provide real support; •to develop new capabilities for applying distributed data processing principles and utilizing high-performance computing architectures in this context; •to develop the ability to adapt solutions based on data-driven approaches for specific problems within a particular field of applicability; •to develop the ability to plan and organize research activities in a field by organizing group work; to understand responsibilities and adhere to professional ethics, as well as data security and confidentiality rules. <p>At the knowledge level:</p>

	<ul style="list-style-type: none"> •to acquire knowledge and competencies to use the technologies and tools derived from the essence and specifics of the "Big Data Technologies" course in data engineering, data science, and business activities; •to develop interdisciplinary competencies by using languages such as Python, platforms like Power BI, Visual Studio, cloud technologies, and cloud platforms such as AWS, Azure, Google Cloud, for Big Data analysis and analytics; •to develop competencies that provide an understanding of business architecture, the impact of introducing new IT technologies, including Big Data technologies, on enterprise management efficiency, and the transformation of the value chain; •to develop management competencies in the field of implementing Big Data systems and services based on Big Data technologies; •to develop and acquire new knowledge and competencies in the use of effective learning, information, and specialization methods and techniques; the ability to select and critically analyze bibliographic resources; the commitment to a continuous professional development plan through the ongoing updating of knowledge and skills in alignment with the scientific and technological advancements in the field. <p>At the skills level:</p> <ul style="list-style-type: none"> •to develop practical skills and abilities in the processes of using technologies and tools specific to the field of "Big Data Technologies" for data engineering, data science, and business activities; •to develop new capabilities to operate with fundamental concepts from mathematical, computer science, and econometric modeling, as well as from statistical analysis, thereby acquiring skills that are applied in practical research or production contexts; •to develop practical skills and acquire new competencies, both in data engineering (the practice of designing and building systems for collecting, storing, and analyzing large-scale data) and analytics (statistical modeling, supervised and unsupervised learning methods, machine learning and deep learning techniques), enabling future specialists to apply the accumulated knowledge to solve diverse problems in the national economy by simulating processes and phenomena in fields relevant to future ICT specialists; •to develop skills in using platforms and technologies specific to processing large volumes of data and implementing scalable applications; •to acquire mathematical and technological knowledge and skills for selecting, evaluating, analyzing, and using Big Data tools and technologies; •to acquire research knowledge and skills in big data analysis, stochastic optimization, predictive modeling, forecasting, enterprise data management, business analysis, and economic and mathematical modeling.
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7. Course/Module content

Syllabus of teaching activities	Number of hours
Course topics	
TPI Introduction to Big Data: definition of Big Data and the reasons for its emergence. Examples of business opportunities and Big Data. Characteristics of Big Data. Types of Big Data. The difference between Big Data and regular data. (distributed structure of Big Data). Managing and analyzing Big Data using microservices technology. Areas of application of Big Data. Advantages of applying Big Data.	2

TP2 Big Data Technologies, TBD: data collection (during the collection process, data is subject to cleaning); data storage (Hadoop, MongoDB, etc., DWH/ETL vs ELT, Data Lake (procedure: extraction, loading into the database/DWH, and transformation into a format suitable for current tasks), DBMS (relational/SQL, MS SQL Server, VS, Power BI, non-relational/NoSQL, MongoDB, Atlas/Compass MongoDB)); data processing (MapReduce technology, Hadoop, Apache Spark); data analysis/processing/analytics (BI technologies/Data analytics/Data Mining, Microsoft Azure).	2
TP3 TBD and limitations of traditional Big Data processing: Data integrity issue. Scalability challenge. ACID transactions. Data scaling and storage across multiple levels: CAP theorem; BASE theorem. NoSQL paradigm. Classification of NoSQL databases.	2
TP4 TBD and high-performance computing: distributed computing across multiple servers, the MapReduce computing paradigm. The Apache Hadoop project and its ecosystem. Apache Spark and its components.	2
TP5 TBD and data repositories: Data management. Security. Data quality (Data discovery, Data auditing, Data provenance, Data exploration). The difference between data lakes and data warehouses.	2
TP6 TBD and Cloud Computing: Cloud service providers - AWS (Amazon), Azure, Google Cloud. Features. Cloud service providers – Data Lake. Definition. Basic concepts of Data Lake. Data Lake architecture. Benefits and risks of using Data Lake in TBD.	2
TP7 TBD and DWH + Data Lake or what is LakeHouse: Components of LakeHouse (Delta Tables, Work Catalog). Delta Lake. Delta Lake architecture. How Delta Lake differs from traditional Big Data processing? Why is Delta Lake important? The need for Delta Lake. Azure Delta Lake. What is ETL and why is it important? And ELT? Differences. Uses. Hybrid Data Lake. Advantages. How do we utilize the advantages of Data Warehouse with Data Lake in TBD?	2
TP8 TBD and Azure Databricks: Introduction to Azure Databricks. Spark architecture. Data manipulation in Azure Databricks. Streaming data processing with Azure Databricks structured streaming. Integration of Azure Databricks with Azure Synapse. Azure Databricks Lakehouse ETL and TBD.	2
TP9 TBD and Big Data Applications: Big Data Analytics. Converting unstructured data into actionable data, knowledge, and wisdom. DWH in Big Data, Big Data management tools in Big Data Analytics.	2
TP10 TBD and the NoSQL paradigm: NoSQL Database Management System. NoSQL vs SQL — which type of database is more efficient for Big Data applications. Tools for storing Big Data.	2
Total lectures:	20

8. Content of the practical works

Syllabus of teaching activities	Number of hours
Laboratory works topics	
LP1. Big Data Analytics – basic stages: Data collection and integration; Data processing; Data cleaning and transformation; Data analysis. Using Big Data Analytics in: Descriptive analysis (information) (“what happened”); Diagnostic analysis (“why it happened”); Predictive analysis (“what will happen next”); Prescriptive analysis (“how to improve it”);	4
LP2. Big Data Analytics technologies and tools, including: for storage, exploitation, analysis, and visualization / Data storage and integration (MongoDB, etc); Data Mining (VS, RapidMiner, ElasticSearch, etc); Big Data Analytics (Spark, Ip. R, Python, etc); Data visualization (Excel, Tableau, MongoDB Charts, Plotly, etc).	4
LP3. Solving challenges in Big Data analysis: (Data storage and processing; Data quality; Data security; Correct tools and techniques) using Atlas MongoDB, Azure, and their numerous easy-to-use technologies. Big Data Analytics in BI and Data Science. Azure Synapse Analytics.	4
LP4. Big Data Analytics applications in business: (Risk management; Understanding market competitors; Identifying trends and patterns; Using Big Data analysis in retail; Creating new products; Managing human resources, in DSS, etc.)	4

LP5. Complex analytical methods: classification of analysis tasks: Text, Data, Web, Social Mining. Applying machine learning in analysis. K-means and C-means clustering, classification. Logistic regression, associations, Apriori algorithm. Text analysis: Search engines: Lucene, Solr, ElasticSearch. Word2Vec and Glove algorithms.	4
Total practical works:	20

9. Using generative AI

Permission to use	<p>The use of generative AI in assignments and projects is permitted, provided that students adhere to the following rules:</p> <ul style="list-style-type: none"> • Generative AI may be used to generate ideas, text structures, or code, but all generated materials must be reviewed and adjusted by the student to ensure that they meet academic requirements. • Any use of generative AI must be declared in the appendix section of each paper, using the phrase: "During the preparation of this paper, the author used [NAME OF TOOL / SERVICE] for the purpose of [REASON]. After using this tool / service, the author reviewed and edited the content as necessary and assumes full responsibility for the content of the paper."
Restrictions to use	<p>Students <i>MUSTN'T consider generative AI as a reliable source of information</i>, as it does not provide clear references or documented sources.</p> <ul style="list-style-type: none"> • <i>Direct citation of AI-generated content</i> in academic papers as if it were a primary source <i>isn't permitted</i>. • Activities in which the use of generative AI is prohibited are specified by the teacher and are usually <i>intermediate and final assessments</i> or that don't involve professional competence development activities.

10. Bibliographic references

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14. **20 Best Big Data Books for Beginners**
<https://bookauthority.org/books/beginner-big-data-books>
15. **Must Read Books for Beginners on Big Data, Hadoop and Apache Spark**
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11. Evaluation

Periodic		Current	Individual study	Project/thesis	Exam
PE 1	PE 2				
15%	15%	15%	15%	-	40%
Minimum performance standards: <ol style="list-style-type: none"> Attendance and activity in lectures and laboratory/practical/seminar work; Achieving the minimum grade of "5" for each of the tests/assessments and individual or group practical works; Demonstrating in the final examination paper the theoretical knowledge, competencies, and practical skills required for the use of Big Data Technologies in the development of a specific project, regarding the Course Final Paper. 					

12. Evaluation criteria

Activity	Evaluation components	Evaluation method, Evaluation criteria	Weight in the final grade of the activity	Weight in the course evaluation
Full-time education				
Periodic evaluation I	Theoretical content, topics 1-3	Test on MOODLE	100%	15%
Periodic evaluation II	Theoretical content, topics 4-5	Test on MOODLE	100%	15%
Current evaluation	Practical activity	Discussions during seminars Test on MOODLE	50%	15%
File completed with Reports for each Case Study under discussion			50%	15%
Individual study	Research on the topic Course final project.	Presentation/public speech	100%	
Final evaluation	Theoretical and practical content	Oral exam. Course final project presentation. Grading according to grading scale	100%	40%