# **COURSE OUTLINE**

# (1) GENERAL

SCHOOL	ENGINEERING				
ACADEMIC UNIT	DEPARTMENT OF COMPUTER ENGINEERING AND INFORMATICS				
LEVEL OF STUDIES	Undergraduate				
COURSE CODE	CEID_24EE594 SEMESTER Spring				
COURSE TITLE	ALGORITHMIC TECHNIQUES FOR DATA SCIENCE				
INDEPEND	ENT TEACHING ACTIVITIE	S			
	its are awarded for separate components of the course,				
e.g. lectures, laboratory exercises, etc. If the credits are awarded			WEEKLY TEACHING HOURS	CREDITS	
for the whole of the course, give the weekly teaching hours and					
	the total credits			_	
	Lectures, Tutorials	s, Laboratory	2(L), 2(T), 1(Lab)	5	
Add rows if necessary. The organisation of teaching and the					
teaching methods used are described in detail at (d).					
COURSE TYPE	Specialized general knowledge, skills development				
general background, special background,					
specialised general					
knowledge, skills					
development					
PREREQUISITE	Recommended prerequisite knowledge: Introduction to Algorithms				
COURSES:	(CEID_23Y205), Data Structures (CEID_23Y210), Linear Algebra (CEID_22Y102),				
	Probability and Statistics (CEID_23Y2O4), or equivalent.				
LANGUAGE OF	Greek (English if there are foreign students)				
INSTRUCTION and					
EXAMINATIONS:					
IS THE COURSE	Yes				
OFFERED TO					
ERASMUS					
STUDENTS					
COURSE WEBSITE	https://www.ceid.upat	tras.gr/webpag	ges/faculty/zaro/teach	ing/atds/index.html	
(URL)					

# (2) LEARNING OUTCOMES

# Learning outcomes

The course learning outcomes, specific knowledge, skills and competences of an appropriate level, which the students will acquire with the successful completion of the course are described.

Consult Appendix A

- Description of the level of learning outcomes for each qualifications cycle, according to the Qualifications Framework of the European Higher Education Area
- Descriptors for Levels 6, 7 & 8 of the European Qualifications Framework for Lifelong Learning and Appendix B
- Guidelines for writing Learning Outcomes

# Upon conclusion of the course the students ought to be able to:

- Understand techniques and applications of fundamental and advanced algorithms for harnessing large data sets.
- Apply algorithmic techniques along with their efficient implementation for analyzing large data sets.
- Conduct extensive experimental methodologies for big data analytics and for empirically assessing algorithms on real-world big data sets.

# Upon conclusion of the course the students are expected to have the following skills/competences:

 Ability to connect data to underlying real-world phenomena and think critically about conclusions drawn from big data analysis.



- Ability to conceive abstractions and processes allowing inferential procedures to be embodied in algorithms for large data sets, ensuring scalability, robustness and understandability.
- Hands-on experimentation with real-world large data sets such as economics data, document collections, geographic data, social and collaboration networks, etc.

#### **General Competences**

Taking into consideration the general competences that the degree-holder must acquire (as these appear in the Diploma Supplement and appear below), at which of the following does the course aim?

Search for, analysis and synthesis of data and information, with the use of the necessary technology

Adapting to new situations

Project planning and management Respect for difference and multiculturalism Respect for the natural environment

Decision-making Showing social, professional and ethical responsibility and

Working independently sensitivity to gender issues
Team work Criticism and self-criticism

Working in an international environment Production of free, creative and inductive thinking Working in an interdisciplinary environment ......

Production of new research ideas Others...

• Search for, analysis and synthesis of data and information, with the use of the necessary

- Adapting to new situations
- Decision-making
- Working independently
- Production of new research ideas
- Criticism and self-criticism
- Production of free, creative and inductive thinking

#### (3) SYLLABUS

- Introduction to programming techniques for large data sets
- Complexity of Algorithms for Large Data Sets
- Exploration of Frequent Itemsets
- Locality-Sensitive Hashing
- Clustering
- Dimensionality Reduction
- Link analysis of large graphs (e.g., PageRank)
- Social Networks and Community Detection
- Efficient Algorithms on Large Graphs
- Recommender Systems
- Algorithms for Large Data Streams

#### (4) TEACHING and LEARNING METHODS - EVALUATION

Face-to-face, Distance learning, etc.	exemplary solutions of exercis	•	
USE OF INFORMATION AND	ICT methods are used in both teaching and communication		
COMMUNICATIONS TECHNOLOGY	with the students. Lecture	slides and supplementary	
Use of ICT in teaching, laboratory education,	material are uploaded in the course's web site.		
communication with students	· ·		
TEACHING METHODS	Activity	Semester workload	
1 1 1 1 1	Activity	Serriester Workload	
The manner and methods of teaching are	Lectures	2x13=26	
The manner and methods of teaching are described in detail.			
The manner and methods of teaching are	Lectures	2x13=26	
The manner and methods of teaching are described in detail. Lectures, seminars, laboratory practice,	Lectures Tutorials (exercises)	2x13=26 2x13=26	



etc.  The student's study hours for each learning	Weekend study Study for exam preparation	2x13=26 6x3=18
activity are given as well as the hours of non-	p specific	
directed study according to the principles of the FCTS	Course total (25-30 hours	148

# STUDENT PERFORMANCE EVALUATION

Description of the evaluation procedure

Language of evaluation, methods of evaluation, summative or conclusive, multiple choice questionnaires, short-answer questions, openended questions, problem solving, written work, essay/report, oral examination, public presentation, laboratory work, clinical examination of patient, art interpretation, other

Specifically-defined evaluation criteria are given, and if and where they are accessible to students

The language of instruction and examination is Greek. Special provisions (lecture notes and examinations in English) can be made for foreign students.

Evaluation (criteria can be found in the web site of the course):

- Programming exercises (30% of final mark), which are compulsory and aim at familiarizing students with the efficient management and analysis of real-world big data sets in practice.
- Final examination (70% of final mark): oral or written examination of graded difficulty, including short-answer questions, algorithm design for problem solving, proofs of algorithm correctness and complexity, as well as problem modelling and solving.

# (5) ATTACHED BIBLIOGRAPHY

#### - Suggested bibliography:

- Jure Leskovec, Anand Rajaraman, Jeffrey D. Ullman: Mining of Massive Datasets, 3rd edition, Cambridge University Press (2019).
- Steven Skienna. The Data Science Design Manual, Springer (2017).
- Avrim Blum, John Hopcroft and Ravindran Kannan: Foundations of Data Science (2017).
- Reza Zafarani, Mohammad Ali Abbasi, Huan Liu: Social Media Mining An Introduction.
   Cambridge University Press (2014)
- Jimmy Lin and Chris Dyer: Data-Intensive Text Processing with MapReduce. Morgan & Claypool Publishers (2010).
- Lecture notes and slides uploaded in the web site of the course.

# - Related academic journals:

- IEEE Transactions on Data Engineering.
- Algorithmica, Springer.
- VLDB Journal
- IEEE Trans. on Pattern Analysis and Machine Intelligence
- Machine Learning, Springer
- Data Mining and Knowledge Discovery
- Data Science Journal
- International Journal of Data Science and Analytics, Springer
- Artificial Intelligence, Elsevier
- Journal of Big Data, Springer
- Big Data Research, Elsevier

### -Related academic conferences:

- IEEE Int'l Conference on Data Engineering
- IEEE Int'l Conference on Data Mining
- Intl. Conf. on Very Large Databases
- ACM SIGIR Conference on Research and Development in Information Retrieval
- ACM Symposium on Principles of Database Systems
- Int'l World Wide Web Conference



- ACM SIGMOD Int'l Conference on Management of Data
- ACM SIGKDD Int'l Conference on Knowledge Discovery and Data Mining
- ACM Int'l Conference on Web Search and Data Mining
- SIAM Int'l Conference on Data Mining
- Int'l Conference on Computational Statistics

