		Teaching	g Guide		
	Identifyin	g Data			2022/23
Subject (*)	Data Analytics with HPC Code			Code	614473108
Study programme	Mestrado Universitario en Compu	ıtación de Altas	Prestacións / Hig	h Performance Comput	ting (Mod. Presencial)
		Descri	ptors		
Cycle	Period	Ye	ar	Туре	Credits
Official Master's Degre	e 2nd four-month period	Fin	st	Optional	6
Language	English				
Teaching method	Face-to-face				
Prerequisites					
Department	Enxeñaría de Computadores				
Coordinador	López Taboada, Guillermo E-mail guillermo.lopez.taboada@udc.es			aboada@udc.es	
Lecturers	López Taboada, Guillermo		E-mail guillermo.lopez.taboada@udc.es		
	Rodríguez Álvarez, Gabriel		gabriel.rodriguez@udc.es		
Web	aula.cesga.es	'			
General description	The increasing amount of informa	tion available th	nrough the Interne	et calls for the efficient p	processing of large amounts of
	data. This has led to the developr	ment of new sto	rage and process	sing techniques to deal v	with huge amounts of data,
	namely Big Data techniques, that naturally adapt to distributed systems.				
	The main goal of this subject is to learn suitable processing techniques for large amounts of information in the Big			of information in the Big Data	
	world, particularly using the Hado	op ecosystem,	and compare the	se techniques with the t	raditional ones employed in HPC
	environments. This will allow the	student to selec	t the optimal tools	s to solve a particular pr	oblem.

	Study programme competences / results
Code	Study programme competences / results
A1	CE1 - Define, evaluate and select the most appropriate architecture and software to solve a problem
A2	CE2 - Analyze and improve the performance of a given architecture or software
B1	CB6 - Possess and understand the knowledge that give a baseline or opportunity to be original in the development and/or application of
	ideas, often in a research environment
B2	CB7 - The students have to know how to apply the acquired knowledge and their capacity to solve problems in new or hardly explored
	environment inside wider contexts (or multidiscipinary) related to its area of development
B6	CG1 - Be able to search and select useful information to solve complex problems, using the bibliographic sources of the field
B8	CG3 - Be able to maintain and extend properly funded theoretical hypothesis to allow the introduction and exploitation of novel and
	advanced technologies in the field
B10	CG5 - Be able to work in teams, specially multidisciplinary, and do a proper time and people management and decision taking
C1	CT1 - Use the basic technologies of the information and computing technology field required for the professional development and the
	long-life learning
C4	CT4 - Value the importance of research, innovation and the technological development in the socioeconomical and cultural advance of the
	society
	long-life learning CT4 - Value the importance of research, innovation and the technological development in the socioeconomical and cultural actions.

Learning outcomes			
Learning outcomes	Stud	y progra	ımme
	con	npetenc	es/
		results	
The student will be capable of installing, configuring, and managing the basic software for massive data processing.	AJ1	BJ2	CJ1
	AJ2	BJ6	
		BJ8	
		BJ10	

The student will be capable of coding massive data processing applications using domain-specific languages.	AJ2	BJ1	CJ1
		BJ2	
		BJ10	
The student will learn about Data Engineering tools (for Intake/Storage/Processing/Visualization).	AJ1	BJ1	CJ1
	AJ2	BJ2	CJ4
The student will learn the skills to search, select and manage Big data-related resources (bibliography, software, etc.).	AJ1	BJ1	CJ1
	AJ2	BJ6	CJ4

	Contents
Topic	Sub-topic
Introduction to Data Engineering	1.1 HPC vs Big Data: similarities and differences in data management.
	1.2 Hardware and Software Technologies for High Performance Data Engineering
	1.3 Data Engineering in HPC infrastructures vs. Cloud environments
2. Introduction to Data Analytics	2.1 Exploratory Data Analytics
	2.2 Introduction to Machine Learning
3. Data Engineering phases	3.1 Modeling (Formats, Compression, Designing Schemas)
	3.2 Intake (Periodicity, Transformations, Tools)
	3.3 Storage (HDFS and NoSQL DBs, HBase, MongoDB, Cassandra)
	3.4 Processing (Batch, Real-Time)
	3.5 Orchestration
	3.6 Analysis (SQL, Machine Learning, Graphs, UI)
	3.7 Governance
	3.8 Integration with BI (Visualization)
4. Use cases	4.1 Applications to Internet of Things (Smart environments and Industry 4.0)
	4.2 Applications to sciences and engineering

	Plannir	ng		
Methodologies / tests	Competencies /	Teaching hours	Student?s personal	Total hours
	Results	(in-person & virtual)	work hours	
Guest lecture / keynote speech	A1 A2 B1 C4	18	0	18
Laboratory practice	B1 B8 B10	20	60	80
Supervised projects	A1 A2 B1 B2 B8	0	45	45
Directed discussion	B6 C1 C4	4	2	6
Personalized attention		1	0	1
(*)The information in the planning table is for	guidance only and does no	t take into account the	heterogeneity of the stud	dents.

	Methodologies
Methodologies	Description
Guest lecture /	Taught by a professor. Classes include theoretical contents, as well as seminars.
keynote speech	
Laboratory practice	Problem solving and practical cases.
Supervised projects	Semi-autonomous work on larger practical cases, under the professors' guidance.
Directed discussion	Guidance to solve individual / group assignments, problem solving and continuous evaluation activities.

Personalized attention	
Methodologies	Description



Directed discussion	During laboratory practice, supervised projects, and directed discussions, students will be able to ask questions, doubts, etc.
Laboratory practice	The teacher, after listening to the students feedback, will go over difficult concepts, solve new problems, or use any
Supervised projects	appropriate methodology to answer the questions.

		Assessment	
Methodologies	Competencies /	Description	Qualification
	Results		
Laboratory practice	B1 B8 B10	Grading the assignments submitted by students.	50
Supervised projects	A1 A2 B1 B2 B8	Grading the supervised projects submitted by students.	50

Assessment comments

Not graded: Students that do not present any practical exercise or guided project will not be graded.

Second opportunity (June/July): Resubmit those laboratory practices or supervised projects not previously presented or submitting improved versions of previously presented practices/projects.

In the case of fraudulent performance of practices or projects the regulations of the University will be applied.

	Sources of information
Basic	- Tom White (2015). Hadoop: The Definitive Guide. O'Reilly (4ª ed.)
	- Wes McKinney (2017). Python for Data Analysis: Data Wrangling with Pandas, NumPy, and IPython. O'Reilly (2ª ed.)
Complementary	- Alex Holmes (2014). Hadoop in practice. Manning (2ª ed.)

Recommendations
Subjects that it is recommended to have taken before
Subjects that are recommended to be taken simultaneously
Subjects that continue the syllabus
Other comments

RecommendationsDue to the large practical component of the subject, it is advisable to be up-to-date with practices and guided projects during the semester.

Observations The course makes intensive use of online communication tools: Video calls, chats, etc. In-person classes will be recorded for later perusing. An online learning management will be using for distributing notes, creating forums, etc.

 $\verb§Anbsp]{The software tools used in this course are generally open-source or have free license for students.}$

(*)The teaching guide is the document in which the URV publishes the information about all its courses. It is a public document and cannot be modified. Only in exceptional cases can it be revised by the competent agent or duly revised so that it is in line with current legislation.