		Teaching Guide			
	Identifyi	ng Data		2020/21	
Subject (*)	Data Analytics with HPC		Code 614973108		
Study programme	Mestrado Universitario en Comp	utación de Altas Prestacións / Hi	Altas Prestacións / High Performance Computing (Mod. Virtual)		
		Descriptors			
Cycle	Period	Year	Туре	Credits	
Official Master's Degree	e 2nd four-month period	First	Optional	6	
Language	English				
Teaching method	Non-attendance				
Prerequisites					
Department	Departamento profesorado mást	erEnxeñaría de Computadores			
Coordinador	López Taboada, Guillermo	E-mail	guillermo.lopez.t	aboada@udc.es	
Lecturers	López Taboada, Guillermo	E-mail	guillermo.lopez.t	aboada@udc.es	
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Web	aula.cesga.es				
General description	The increasing amount of inform	ation available through the Intern	et calls for the efficient p	processing of large amounts of	
	data. This has led to the develop	ment of new storage and proces	sing techniques to deal v	with huge amounts of data,	
	namely Big Data techniques, tha	t naturally adapt to distributed sy	stems.		
	The main goal of this subject is to	o learn suitable processing techn	iques for large amounts	of information in the Big Data	
	world, particularly using the Hade	oop ecosystem, and compare the	ese techniques with the t	raditional ones employed in HPC	
	environments. This will allow the	student to select the optimal too	ls to solve a particular pr	oblem.	
Contingency plan	1. Modifications to the contents				
	- No changes will be made.				
	2. Methodologies				
	*Teaching methodologies that ar	e maintained			
	- All.				
	3. Mechanisms for personalized	attention to students			
	- Email: Daily. Of use to make co	onsultations, request virtual meeti	ings to resolve doubts ar	nd follow up on supervised work.	
	- CESGA classroom: Daily. Acco	rding to the needs of the student	s. They have "thematic f	orums associated with the	
	modules" of the subject, to formu	late the necessary queries. Ther	e are also ?specific activ	vity forums? to develop the	
	?Directed Discussions?, through	which the development of theore	etical content of the subj	ect is put into practice.	
	- Teams or the Slack + Jitsi com	hination: 1 weekly session in a la		cement of the theoretical contents	
	rearris of the Glack Foliar com	omation. I weekly session in a la	rge group for the advanc	omone or the theoretical content	
	and the tutored works in the time				
	and the tutored works in the time	slot assigned to the subject in the	ne faculty class calendar		
	and the tutored works in the time From 1 to 2 weekly sessions (or	slot assigned to the subject in the subject in the subject in the more as the students demand) in	ne faculty class calendar n a small group (up to 6 p	people), for follow-up and support	
	and the tutored works in the time From 1 to 2 weekly sessions (or	slot assigned to the subject in the more as the students demand) in ork". This dynamic allows for sta	ne faculty class calendar n a small group (up to 6 p	people), for follow-up and support	
	and the tutored works in the time From 1 to 2 weekly sessions (or in carrying out the "supervised w	e slot assigned to the subject in the more as the students demand) in ork". This dynamic allows for star the subject.	ne faculty class calendar n a small group (up to 6 p	people), for follow-up and support	
	and the tutored works in the time From 1 to 2 weekly sessions (or in carrying out the "supervised w students to carry out the work of	e slot assigned to the subject in the more as the students demand) in ork". This dynamic allows for star the subject.	ne faculty class calendar n a small group (up to 6 p	people), for follow-up and support	
	and the tutored works in the time From 1 to 2 weekly sessions (or in carrying out the "supervised w students to carry out the work of 4. Modifications in the evaluation	slot assigned to the subject in the more as the students demand) in ork". This dynamic allows for star the subject.	ne faculty class calendar n a small group (up to 6 p		

Study programme competences	
Code	Study programme competences

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

Learning outcomes			
Learning outcomes	Study	y progra	amme
	COI	mpeten	ces
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
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

Planning	
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Methodologies / tests	Competencies	Ordinary class	Student?s personal	Total hours
		hours	work hours	
Workbook	A1 A2 B1 B6 C4	0	18	18
Laboratory practice	B1 B8 B10	0	80	80
Supervised projects	A1 A2 B1 B2 B8	0	45	45
Directed discussion	B6 C1 C4	4	2	6
Personalized attention		1	0	1
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(*)The information in the planning table is for guidance only and does not take into account the heterogeneity of the students.

	Methodologies	
Methodologies	Description	
Workbook	kbook Planned instruction through various teaching materials.	
Laboratory practice	Laboratory practice Problem solving and practical cases.	
Supervised projects	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	
Laboratory practice	During laboratory practice, supervised projects, and directed discussions, students will be able to ask questions, doubts, etc.	
Supervised projects	The teacher, after listening to the students feedback, will go over difficult concepts, solve new problems, or use any	
Directed discussion	appropriate methodology to answer the questions.	

		Assessment	
Methodologies	Competencies	Description	Qualification
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 (2 ^a 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. 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.