



POLITÉCNICA

INTERNATIONAL  
CAMPUS OF  
EXCELLENCE

COORDINATION PROCESS OF  
LEARNING ACTIVITIES  
PR/CL/001



E.T.S. de Ingenieros  
Informáticos

# ANX-PR/CL/001-01

## LEARNING GUIDE

### SUBJECT

**103000825 - Statistical Data Analysis**

### DEGREE PROGRAMME

10AZ - Master Universitario En Innovación Digital

### ACADEMIC YEAR & SEMESTER

2024/25 - Semester 1

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## 1. Description

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### 1.1. Subject details

<b>Name of the subject</b>	103000825 - Statistical Data Analysis
<b>No of credits</b>	4.5 ECTS
<b>Type</b>	Optional
<b>Academic year of the programme</b>	First year
<b>Semester of tuition</b>	Semester 1
<b>Tuition period</b>	September-January
<b>Tuition languages</b>	English
<b>Degree programme</b>	10AZ - Master Universitario en Innovación Digital
<b>Centre</b>	10 - Escuela Tecnica Superior De Ingenieros Informaticos
<b>Academic year</b>	2024-25

## 2. Faculty

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### 2.1. Faculty members with subject teaching role

<b>Name and surname</b>	<b>Office/Room</b>	<b>Email</b>	<b>Tutoring hours *</b>
Jacinto Gonzalez Pachon	2106	jacinto.gonzalez.pachon@upm.es	Sin horario. Check office hours in September
Arminda Moreno Diaz (Subject coordinator)	2204	arminda.moreno@upm.es	Sin horario. Check office hours in September

Juan Antonio Fdez Del Pozo De Salamanca	2101	juan.fdezpozo.salamanca@u pm.es	Sin horario. Check office hours in September
Antonio Jimenez Martin	2105	antonio.jimenez@upm.es	Sin horario. Check office hours in September

\* The tutoring schedule is indicative and subject to possible changes. Please check tutoring times with the faculty member in charge.

### 3. Prior knowledge recommended to take the subject

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#### 3.1. Recommended (passed) subjects

The subject - recommended (passed), are not defined.

#### 3.2. Other recommended learning outcomes

- Basic knowledge of statistics and inference will be helpful.

### 4. Skills and learning outcomes \*

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#### 4.1. Skills to be learned

CB07 - Que los estudiantes sepan aplicar los conocimientos adquiridos y su capacidad de resolución de problemas en entornos nuevos o poco conocidos dentro de contextos más amplios (o multidisciplinares) relacionados con su área de estudio

CB10 - Que los estudiantes posean las habilidades de aprendizaje que les permitan continuar estudiando de un modo que habrá de ser en gran medida autodirigido o autónomo.

CE-HMDA02 - Capacidad para aplicar técnicas para la generación de visualizaciones adecuadas para el análisis y la exploración de datos en un contexto médico, y para la correcta comunicación de los resultados del análisis

CE-HMDA04 - Capacidad para aplicar métodos avanzados para clasificación, modelado, segmentación y predicción a partir de un conjunto de datos

CG03 - La capacidad de usar la lengua inglesa de manera competente, es decir, con capacitación para tareas

complejas de trabajo y estudio.

## 4.2. Learning outcomes

RA61 - To know and apply dimensionality reduction and modelization techniques to multivariate data.

RA60 - To know and apply the main techniques to explore, describe and analyse multivariate data.

RA62 - To apply the proper methodology to analyse time series.

\* The Learning Guides should reflect the Skills and Learning Outcomes in the same way as indicated in the Degree Verification Memory. For this reason, they have not been translated into English and appear in Spanish.

## 5. Brief description of the subject and syllabus

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### 5.1. Brief description of the subject

The course is intended to be a non-exhaustive survey of techniques to convert multivariate data into useful information so that good decisions can be made. The perspective is twofold, theoretical and applied, covering topics such as: exploratory data analysis, statistical summaries and graphical representations, dimensionality reduction, regression techniques and time series analysis. There will be an emphasis on hands-on application of the theory and methods throughout, with extensive use of R. The course is taught in four modules which are broken down in topics below and properly referenced in the timeline.

### 5.2. Syllabus

1. Using data to answer statistical questions.
  - 1.1. Gathering data. Experiments and observational studies.
  - 1.2. Review of univariate inference. Bootstrap and confidence intervals. Permutation tests.
  - 1.3. Methods for comparing two groups.
2. Exploratory data analysis and descriptive statistics.
  - 2.1. Aspects of multivariate data: exploration and visualization.
  - 2.2. Aspects of multivariate data: descriptive statistics.
  - 2.3. Aspects of multivariate data: dimensionality reduction.
3. Statistical modeling.

- 3.1. Simple Linear Regression.
- 3.2. Multiple Linear Regression.
- 3.3. The General Linear Model.
- 4. Time Series.
  - 4.1. Definitions, Applications and Techniques.
  - 4.2. Stationarity and Seasonality.
  - 4.3. Common approaches.
  - 4.4. Box-Jenkins model identification, estimation and validation.
  - 4.5. Forecasting.

## 6. Schedule

### 6.1. Subject schedule\*

Week	Type 1 activities	Type 2 activities	Distant / On-line	Assessment activities
1	<b>Module 1: Lecture</b> Duration: 02:00 Lecture			
2	<b>Module 1: Lecture</b> Duration: 01:00 Lecture  <b>Computer Lab. Module 1.</b> Duration: 01:00 Additional activities			
3	<b>Module 1: Lecture</b> Duration: 02:00 Lecture			
4	<b>Module 2: Lecture</b> Duration: 02:00 Lecture			
5	<b>Module 2: Lecture</b> Duration: 01:00 Lecture  <b>Computer Lab. Module 2.</b> Duration: 01:00 Additional activities			
6	<b>Module 2: Lecture</b> Duration: 01:00 Lecture  <b>Computer Lab. Module 2.</b> Duration: 01:00 Additional activities			
7	<b>Module 3: Lecture</b> Duration: 02:00 Lecture			
8	<b>Module 3</b> Duration: 02:00 Lecture  <b>Exam on Modules 1 and 2. Progressive evaluation</b> Duration: 02:00 Additional activities			<b>Exam on Modules 1 and 2. Progressive evaluation</b> Written test Progressive assessment Presential Duration: 02:00
9	<b>Module 3: Lecture</b> Duration: 02:00 Lecture			

10	<b>Module 3: Lecture</b> Duration: 02:00 Lecture			
11	<b>Module 4: Lecture</b> Duration: 02:00 Lecture			
12	<b>Module 4: Lecture</b> Duration: 02:00 Lecture			
13	<b>Module 4: Lecture</b> Duration: 01:00 Lecture  <b>Computer Lab. Module 4.</b> Duration: 01:00 Additional activities			
14	<b>Module 4: Lecture</b> Duration: 01:00 Lecture  <b>Computer Lab. Module 4.</b> Duration: 01:00 Additional activities			
15	<b>Module 4: Lecture</b> Duration: 02:00 Lecture  <b>Exam on Modules 3 and 4. Progressive evaluation</b> Duration: 02:00 Additional activities			<b>Exam on Modules 3 and 4. Progressive evaluation</b> Written test Progressive assessment Presential Duration: 02:00
16				<b>Global Exam: 4 exams, one corresponding to each module. Only for students that haven't passed the course yet.</b> Written test Global examination Presential Duration: 03:00
17				

Depending on the programme study plan, total values will be calculated according to the ECTS credit unit as 26/27 hours of student face-to-face contact and independent study time.



## 7. Activities and assessment criteria

### 7.1. Assessment activities

#### 7.1.1. Assessment

Week	Description	Modality	Type	Duration	Weight	Minimum grade	Evaluated skills
8	Exam on Modules 1 and 2. Progressive evaluation	Written test	Face-to-face	02:00	50%	3 / 10	CB07 CG03 CE-HMDA02
15	Exam on Modules 3 and 4. Progressive evaluation	Written test	Face-to-face	02:00	50%	3 / 10	CB07 CB10 CG03 CE-HMDA04

#### 7.1.2. Global examination

Week	Description	Modality	Type	Duration	Weight	Minimum grade	Evaluated skills
16	Global Exam: 4 exams, one corresponding to each module. Only for students that haven't passed the course yet.	Written test	Face-to-face	03:00	100%	3 / 10	CB07 CB10 CG03 CE-HMDA02 CE-HMDA04

#### 7.1.3. Referred (re-sit) examination

Description	Modality	Type	Duration	Weight	Minimum grade	Evaluated skills
Extraordinary exam: all 4 exams, one corresponding to each module.	Written test	Face-to-face	03:00	100%	3 / 10	CB07 CB10 CG03 CE-HMDA02 CE-HMDA04

## 7.2. Assessment criteria

### Progressive evaluation:

The overall course grade will be based on the performance on 4 exams, equally weighed, corresponding to each of the modules taught. The weeks in which each exam is scheduled might change depending on the academic calendar.

Each exam should obtain a grade equal or greater than 3 (0-10 scale) for the overall course grade to be computed. The overall course grade will be the average of the grades obtained in the 4 exams and it must be equal or greater than 5 to pass the course.

No makeup exams will be scheduled.

### January exam:

As far as exams are concerned, the final exam in January is only for those failing to take the midterms when requested or those obtaining an overall grade less than 5. In the latter case, only those specific exams not previously passed have to be taken again.

### Extraordinary exam:

In case you are opting for the extraordinary exam in July's session, you have to warn the instructor(s) in advance (at least two weeks before the exam, by email, but the sooner the better). If you have taken some of the exams on the ordinary session and obtained a grade greater or equal than 3, these grades can be kept and no repetition of these exams is required. Otherwise, all 4 exams can be taken in this opportunity.

## 8. Teaching resources

### 8.1. Teaching resources for the subject

Name	Type	Notes
Moodle	Web resource	Main communication channel with students. Repository: slides, scripts, data sets, other resources.
Johnson, R.A., Whichern, D.W. (2007) Applied Multivariate Statistical Analysis. Pearson Education.	Bibliography	Friendly exposition of the most important multivariate techniques, including clustering. They also introduce other Artificial Intelligence techniques like neural networks.
Rencher, A.C. Methods of Multivariate Analysis.	Bibliography	Clear exposition of Multivariate Analysis Techniques, from a statistical point of view. Many examples.
Everitt, B.S. and Dunn G. (1997) Applied Multivariate Data Analysis. Arnold.	Bibliography	Excellent exposition of multivariate techniques. They make the Generalised Linear Model easily understandable.
Hair, J.F., Black, W.C., Babin, B.J., Anderson R.E. Multivariate Data Analysis.	Bibliography	A Global Perspective on multivariate Techniques. Very detailed examples. In almost every topic, a "Rules of Thumb" section summarizes the relevant facts.
Sharma, S (1996). Applied Multivariate Techniques. Wiley.	Bibliography	
Multivariate Analysis of Ecological Data. Greenacre, Primicerio. Fundación BBVA.	Web resource	<a href="http://www.fbbva.es/TLFU/tlfu/esp/publicaciones/libros/fichalibro/index.jsp?codigo=769">http://www.fbbva.es/TLFU/tlfu/esp/publicaciones/libros/fichalibro/index.jsp?codigo=769</a>   Excellent compilation and description of multivariate techniques applied to Ecological Data. Clear, educational, didactic explanations. Lots of examples

Biplots in Practice. Michael Greenacre. Fundación BBVA.	Web resource	<a href="http://www.multivariatestatistics.org/biplots.html">http://www.multivariatestatistics.org/biplots.html</a> Excellent Monograph on Biplots.
Rawlings, J.O., Pantula, S.G., Dickey, D.A. Applied Regression Analysis.	Bibliography	Almost everything about Regression Models.
Chatfield, C. (2003) The Analysis of Time Series: An Introduction. Chapman and Hall.	Bibliography	
Robert Hyndman, George Athanasopoulos. Forecasting: principles and practice. OTexts, 2018.	Web resource	Bibliography and web resource. Available at <a href="https://otexts.com/fpp2/">https://otexts.com/fpp2/</a>
Robert Nau. Statistical forecasting: notes on regression and time series analysis.	Web resource	Excelent set of tips to fit ARIMA models. Available at <a href="https://people.duke.edu/~rnau/411home.htm">https://people.duke.edu/~rnau/411home.htm</a>

## 9. Other information

### 9.1. Other information about the subject

Regarding the Sustainable Developing Goals, broadly speaking, proper data analysis leads to an understanding of processes and the generation of information available to everyone. This information can help clarify and make more transparent the way in which institutions, for instance, try to meet people's needs. This information increases transparency, accountability and citizen participation in institutions and facilitates information-sharing between different actors. In this particular case, proper analysis of data might help in achieving SDG 16: Promote peaceful and inclusive societies for sustainable development, provide access to justice for all and build effective, accountable and inclusive institutions at all levels.