

Longitudinal Data Analysis

Code: 104879
ECTS Credits: 6

Degree	Type	Year	Semester
2503852 Applied Statistics	OT	4	0

The proposed teaching and assessment methodology that appear in the guide may be subject to changes as a result of the restrictions to face-to-face class attendance imposed by the health authorities.

Contact

Name: Juan Ramón González Ruíz
Email: JuanRamon.Gonzalez@uab.cat

Use of Languages

Principal working language: spanish (spa)
Some groups entirely in English: No
Some groups entirely in Catalan: No
Some groups entirely in Spanish: No

Prerequisites

- It is recommended, but not essential because a class will be held to homogenize the level of the students, to know the generalized linear models and the Cox model for the survival analysis.
- The subject on observational studies contains a complementary introduction to the topic that will deal with the analysis of the temporal evolution of incidence and mortality rates in which the concept of rate will be introduced.
- It is advisable but not essential to know the statistical package R.

Objectives and Contextualisation

The main objectives are:

- To know the statistical models for the analysis of longitudinal data (information obtained from measurements made over time) that often appear in health sciences (biology, medicine, pharmacology, toxicology, chemistry and / or engineering))
- To know the statistical models to analyze the temporal evolution of the incidence and mortality rates of a disease to detect temporary changes and why they are due
- To know the statistical models to analyze the time until the occurrence of an event of interest that appears recurrent (tumor relapse, migraine, heart attacks, ...) taking into account the effect of covariates, the effect of the intervention and / or the effect to observe several events beforehand
- To know the statistical models to analyze data obtained from repeated measurements over time using linear models (rehospitalizations, relapse of a disease, ...)
- To know the statistical models to analyze data obtained from repeated measurements over time using non-linear models (tumor growth in rats, evolution of the weight of children after birth, ...)
- To be able to read critically a scientific article that considers the analysis of a study in which information collected over time is available.

- Be able to identify the statistical model necessary to analyze a set of data that will be presented in practical exercises and that belong to real studies.
- Know how to perform all these analyzes using R using the appropriate libraries.

Competences

- Correctly use a wide range of statistical software and programming languages, choosing the best one for each analysis, and adapting it to new necessities.
- Critically and rigorously assess one's own work as well as that of others.
- Formulate statistical hypotheses and develop strategies to confirm or refute them.
- Identify the usefulness of statistics in different areas of knowledge and apply it correctly in order to obtain relevant conclusions.
- Interpret results, draw conclusions and write up technical reports in the field of statistics.
- Make efficient use of the literature and digital resources to obtain information.
- Students must be capable of applying their knowledge to their work or vocation in a professional way and they should have building arguments and problem resolution skills within their area of study.
- Students must be capable of collecting and interpreting relevant data (usually within their area of study) in order to make statements that reflect social, scientific or ethical relevant issues.
- Students must be capable of communicating information, ideas, problems and solutions to both specialised and non-specialised audiences.
- Use quality criteria to critically assess the work done.
- Work cooperatively in a multidisciplinary context, respecting the roles of the different members of the team.

Learning Outcomes

1. Critically assess the work done on the basis of quality criteria.
2. Design and conduct hypothesis tests in the different fields of application studied.
3. Draw conclusions that are consistent with the experimental context specific to the discipline, based on the results obtained.
4. Draw up technical reports that clearly express the results and conclusions of the study using vocabulary specific to the field of application.
5. Interpret statistical results in applied contexts.
6. Justify the choice of method for each particular application context.
7. Make effective use of references and electronic resources to obtain information.
8. Reappraise one's own ideas and those of others through rigorous, critical reflection.
9. Recognize the importance of the statistical methods studied within each particular application.
10. Students must be capable of applying their knowledge to their work or vocation in a professional way and they should have building arguments and problem resolution skills within their area of study.
11. Students must be capable of collecting and interpreting relevant data (usually within their area of study) in order to make statements that reflect social, scientific or ethical relevant issues.
12. Students must be capable of communicating information, ideas, problems and solutions to both specialised and non-specialised audiences.
13. Use different programmes, both open-source and commercial, associated with the different applied branches.
14. Work cooperatively in a multidisciplinary context, accepting and respecting the roles of the different team members.

Content

These are the proposed contents*:

1. Introduction to the course
 - 1.1 Introduction to R markdown: creation of automated and reproducible reports
 - 1.2 Tidyverse

2. Analysis of incidence and mortality rates
 - a. Introduction
 - b. Rate definition
 - c. Calculation of standardized rates
 - d. Analysis of temporal trends
 - e. Regression models 'jointpoint'
 - f. Age-period-cohort models

3. Survival analysis for data with recurrent events
 - a. Introduction
 - b. Non-parametric models
 - i. Model Peña-Strawderman-Hollander
 - ii. Model of Chan-Wang
 - iii. Fragility model ('frailty model')
 - c. Semi-parametric models
 - i. Conditional model (Prentice-William-Peterson)
 - ii. Marginal model (Wei-Lin-Weidsfeld)
 - iii. Fragility model
 - iv. General model (Peña-Hollander)
 - v. Cancer model (González-Peña-Slate)
 - d. Model with terminal event
 - i. Estimation through penalized credibility

4. Analysis of longitudinal data by linear models
 - a. Introduction
 - b. Designs with repeated measures
 - c. Repeated measures ANOVA
 - d. MANOVA
 - e. Mixed linear model.
 - f. Diagnosis of the model

5. Analysis of longitudinal data through non-linear models
 - a. Introduction
 - b. Graphical inspection of the data
 - c. Estimation of a non-linear model
 - d. Diagnosis of the model
 - e. Solutions when model assumptions are not met
 - f. Model selection
 - g. Nonlinear Mixed model

*Unless the requirements enforced by the health authorities demand a prioritization or reduction of these contents.

Methodology

Theoretical sessions:

In these sessions the main concepts of each topic will be presented, as well as the analysis of data with real examples in which the necessary code of R will be displayed to carry out this task. The slides (created with R Markdown - which guarantees reproducibility of the results) will include the theoretical concepts, the data analysis and the interpretation of the results and the conclusions drawn from them.

Practical sessions:

In these sessions some guided exercises will be proposed that the student will have to solve individually. Each student will have an individual data set for the same problem. A random database for the same real problem will be generated independently for each student. With this methodology, the student is expected to investigate

and learn how to analyze a set of real data, regardless of whether he asks a classmate how to carry it out, since each student will have to analyze their data and obtain their conclusions from their own results.

Assistance to seminars:

Exceptionally, if there are certain time matches and if the students see it feasible to do it, the students, together with the teacher, will attend a seminar organized by the UAB

Applied Statistics Service or by another research center close to the College. This assistance will not be compulsory, but it will be highly recommended by the teacher, since the students could see how the methodology they are learning is used in real studies and they could see how the work of a statistician has a crucial implication in the completion of said investigations.

Individualized work

The solutions of all the practices that are carried out in person must be delivered to the teacher. As already mentioned, these practices will be carried out individually, as each student will have a personalized database on the same problem to be addressed. In addition to these face-to-face practices, the student will have to solve four other practices at home and deliver the numerical solution, as well as the R code used to obtain said results. Both the classroom practices and those carried out at home will be part of the continuous evaluation. The student will have access to all the teaching material from a teaching web in which there will also be a forum in which to raise questions that could ideally be resolved by their peers and that the teacher will supervise and / or solve, if necessary.

PS: The proposed teaching methodology may experience some modifications depending on the restrictions to face-to-face activities enforced by health authorities.

Annotation: Within the schedule set by the centre or degree programme, 15 minutes of one class will be reserved for students to evaluate their lecturers and their courses or modules through questionnaires.

Activities

Title	Hours	ECTS	Learning Outcomes
Type: Directed			
Theoretical sessions	150	6	8, 1, 2, 4, 3, 5, 6, 12, 10, 11, 9, 14, 13, 7

Assessment

Self-evaluation:

After each theoretical session, a series of general questions will be proposed to evaluate if the student has learned the basic concepts of the topic treated during that session.

Delivery of practices:

During the course the student will have to solve several practices in which the teacher will be able to guide since they will be done in person. The numerical solution must be sent as well as the R code used to solve them.

Final Exam:

Students will take a face-to-face test to assess whether they have acquired the minimum theoretical and practical concepts about the subject. This test will contain conceptual questions about the models treated in class and R outputs similar to those obtained in the analyzes that have been made during the course, on which questions will be asked to interpret the results.

Final mark will be:

$$0,5 \cdot NE + 0,5 \cdot NPA$$

where NE stands for the exam and NPA for practical and self-evaluation marks, respectively. It is mandatory to have $NE \geq 5$ to pass the topic.

PS: Student's assessment may experience some modifications depending on the restrictions to face-to-face activities enforced by health authorities.

Assessment Activities

Title	Weighting	Hours	ECTS	Learning Outcomes
Individual tasks	50%	0	0	8, 1, 2, 4, 5, 6, 12, 10, 11, 14, 13, 7
Test	50%	0	0	3, 5, 6, 9

Bibliography

González JR, Llorca F, Moreno V. Algunos aspectos metodológicos sobre los modelos edad-periodo-cohorte. Aplicación a las tasas de mortalidad por cáncer. Gaceta Sanitaria, 2002;16:267-273

Kim HJ, Fay MP, Feuer EJ, Midthune DN. Permutation tests for joinpoint regression with applications to cancer rates. Statistics in Medicine, 2000;19:335-51

Fernandez E, Gonzalez JR, JM Borrás, et al. Recent decline in cancer mortality in Catalonia (Spain). A Joint point regression analysis. European Journal of Cancer, 2001;37:2222-2228.

Gonzalez JR, Peña E, Slate E. Modelling intervention effects alter cancer relapses. Statistics in Medicine, 2005;24:3959-1975

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Gonzalez JR. Modelling recurrent event data with application to cancer research. VDM Verlag, Saarbrken, Germany, 2009 (pdf del libro accesible en la web de la asignatura)

Therneau T and Grambsch P. Modeling Survival Data: Extending the Cox Model. Springer-Verlag, New York, 2000.

Software

Theroy and practical lectures will be in R.