Syllabus: Introduction to Social Data Science

Course Information

• Course Code: ULI2008.1 – Sosyal Veri Bilimine Giriş / PSIR3066 – Introduction to Data Science

• Schedule: Monday, 3 hours (14:00–17:00)

• Classroom: Faculty of Political Science, Room RTE.S1.124

• Office Hours: Monday, 09:00–11:00

• Office Location: Faculty of Political Science, Room RTE.S1.131

• Credits & ECTS: 3 credits, ECTS: 5

• Course Type: Elective

• Instructor: Assoc. Prof. Hakan Mehmetcik

• Contact: hakan.mehmetcik@marmara.edu.tr

Course Objectives

This course provides an introduction to **R**, a free and open-source statistical computing environment and programming language widely used in **data analysis**, **statistical research**, **and industrial applications**. The primary goal of this course is to equip students with the ability to use **R** for problem-solving and practical applications in social data science. Students will gain hands-on experience with data manipulation, visualization, statistical computation, and analysis beyond basic spreadsheet tools such as Excel.

R is a versatile programming language that enables efficient data processing, statistical modeling, and visualization. It runs on multiple platforms, including UNIX, Linux, Windows, and macOS.

This course is designed for students who want to learn the fundamentals of R and its applications in social data science. The content focuses on:

- \bullet The ${\bf basics}$ of the R ${\bf environment}$ and how to navigate it effectively.
- Arithmetic and logical operators, data structures, variables, and data types in R.

- Data manipulation and transformation techniques, including working with R packages for visualization, reporting, and statistical analysis.
- Importing, processing, and transforming datasets for analytical purposes.
- Core data science concepts, including parametric and non-parametric statistical tests.

By the end of the course, students will have a strong foundation in **R programming** and its **applications in data science**, enabling them to analyze and visualize real-world datasets effectively.

Course Learning Outcomes

By the end of this course, students will be able to:

- Use R and RStudio effectively for data analysis and programming.
- Process and analyze data using R's built-in functions and libraries.
- Create visualizations and reports to communicate data-driven insights.
- Apply statistical methods for data exploration and hypothesis testing.

These skills will provide students with a strong foundation in data analysis and statistical computing, preparing them for more advanced data science and problem-solving tasks.

Course Readings

Primary Textbooks (Required Readings)

Modern Data Science with R

R for Data Science

Supplementary Resources (Optional Readings & References)

- An Introduction to R (pdf) by W. N. Venables, D. M. Smith, and the R Development Core Team
- Advanced R (2nd Ed) by Hadley Wickham (advanced)
- Intro to R video lectures by Google Developers
- R Programming wikibook
- Using R for Data Analysis and Graphics by J. H. Maindonald
- The R Inferno by Patrick Burns (advanced)

Required Materials for the Course

Hardware Requirement

• Students **must bring a laptop** to class for hands-on exercises.

Software Requirement

To participate in this course, students need to install **R** and **RStudio** on their personal computers.

1. Download and Install R

• First, download and install R from The R Project.

2. Download and Install RStudio

• After installing R, download and install **RStudio** from Posit.

3. Install Required R Packages

 During the course, students will need to install and use the tidyverse and RMarkdown packages. Instructions for installing these and other required packages will be provided in class.

Getting Help

If you need assistance, consider the following steps:

1. Self-Help Resources

- Use R Help (?command) within RStudio.
- Search for solutions on Google or AI platforms (e.g., ChatGPT, DeepSeek).

2. Ask Questions in Google Classroom

- Post your questions in the **Google Classroom discussion board** for peer and instructor support.
- If you know the answer to a classmate's question, feel free to help!

3. Schedule a Virtual Meeting

• If additional support is needed, students can schedule a virtual meeting via Google Meet with the instructor.

Attendance and Participation Requirements

Research indicates that **regular attendance** is one of the strongest predictors of academic success.

According to the university's **Education and Teaching Regulations**, students are required to:

- Attend at least 70% of theoretical classes.
- Attend at least 80% of practical sessions (if applicable).

Please consider these attendance requirements carefully before enrolling in this course!

Grading & Final Project Requirements

Grading Breakdown

• Midterm Project – Phase 1: 40%

• Final Project – Phase 2: 50%

• Participation: 10%

For exact deadlines, please check the university's online system (OBYS).

Each student is required to conduct a **data analysis using R** based on a dataset sourced from **Data is Plural**. The project consists of two phases:

Midterm Project - Phase 1: (Problem Definition & Data Collection)

- 1. Problem Definition & Data Selection:
 - Identify a social issue you want to analyze.
 - Formulate a **clear research question** and determine the relevant dataset.
 - Data must be selected from Data is Plural's dataset archive.
- 2. Data Cleaning & Exploratory Data Analysis (EDA):
 - Prepare the dataset for analysis by handling missing values, outliers, and irrelevant data.
 - Conduct **Exploratory Data Analysis (EDA)** to identify patterns and relationships in the data.

Final Project - Phase 2: (Data Modeling & Reporting)

3. Data Modeling:

- Based on EDA insights, select appropriate statistical or machine learning models to answer your research question.
- Justify your model choice and interpret the results.

4. Presentation & Report Writing:

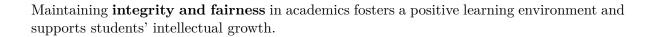
- Clearly present your findings through visualizations, a written report, or a presentation.
- The report should include all project phases, from **problem definition to con- clusions**.

For exact submission deadlines, please check Marmara University's BYS system.

Academic Integrity & Ethical Guidelines

To ensure a fair, transparent, and ethical learning environment, students must adhere to the following principles:

- 1. **Academic Honesty** Plagiarism and cheating are serious offenses and will not be tolerated. Any violations may result in disciplinary action.
- 2. **Exam Conduct** Students must follow all exam rules and refrain from using unauthorized materials (e.g., books, notes, electronic devices).
- 3. **Original Work** All assignments and exams must reflect the student's own work. Proper citation is required for any external sources.
- 4. **Respect for Peers** Students should engage respectfully with classmates and avoid unauthorized collaboration on assignments.
- 5. **Instructor Communication** If you have concerns about academic integrity or ethical dilemmas, discuss them openly with the instructor.
- 6. **Accountability** Students are responsible for their own learning, participation, and adherence to ethical guidelines.
- 7. Consequences of Violations Breaching these principles may result in academic penalties, such as assignment failure or disciplinary actions by the university.



Accessing Google Classroom

To join the course on **Google Classroom**, follow these steps:

- 1. **Sign in to your Google Account** and use the invitation link: Join Google Classroom
- 2. Alternatively, you can manually join:
 - Go to Google Classroom.
 - Sign in with your Google Account.
 - On the "My Classes" page, click "Join".
 - Enter the class code: xweaxw6 and click "Join".

Important Notes:

- Ensure that you enter the **correct class code** to avoid errors.
- Once you join, you will have access to assignments, announcements, and course materials.
- You can also access Google Classroom from your mobile device.
- Make sure you are using the latest version of the Google Classroom app for a smooth experience.

Week	Date	Type	Topic	Readings
1	February 17,		Introduction to R, RStudio, and	MDSR Appendix A-B-C,
	2025	Lesson	R Markdown	MDSR 1, R4DS 1
	February 24,		Data Visualization with ggplot	MDSR 2, R4DS 3
	2025	Lesson		
	March 03,		Data Visualization with ggplot2	MDSR 3, R4DS 3
	2025	Lesson		
4	March 10,		Data Frames and Data	MDSR 4-5, R4DS 5-8
	2025	Lesson	Manipulation	
	March 17,		Data Summarization and Pipes	MDSR 6-7, R4DS 9-16
	2025	Lesson		
	March 24,		Data Communication, Data	MDSR Appendix D,
	2025	Lesson	Import and Workflows	R4DS 26
	April 07,		Data Communication, Data	MDSR Appendix D,
	2025	Lesson	Import and Workflows	R4DS 267
	April 21,		Linear Regression, Inference and	MDSR Appendix E,
	2025	Lesson	Modeling	R4DS 24
	April 28,		Descriptive Statistics, Inference	MDSR 9-10, R4DS 22-23
	2025	Lesson	and Modeling	
	May 05,	Exam	NA	NA
	2025			
	May 12,		Machine Learning, Modeling	MDSR 10-11, R4DS 25
	2025	Lesson		
	May 19,		Modeling, Text as Data	MDSR 19, R4DS 14
	2025	Lesson		
	May 26,	_	Project Sample	Hands-out
	2025	Lesson		
	June 02,	_	Project Sample	Hands-out
	2025	Lesson		
	June 16,	_	Project Sample	Hands-out
	2025	Lesson	<i>T</i>	37.4
	June 23,		Review	NA
	2025	Lesson	27.4	27.4
	June 30,	Exam	NA	NA
	2025			

Readings

Week Date

Type

Topic

Additional Information

- The course **content and structure** may be modified at the instructor's discretion. Any changes will be communicated in advance.
- Students are encouraged to reach out to the **instructor or teaching assistant** if they have any questions or concerns regarding the course. Contact details are provided above.
- Note: This syllabus outlines the **general structure and teaching methods** of the course. Additional course materials and a detailed schedule will be provided as the course progresses.

Wishing you a successful semester!