

## ANALYTICAL TOOLS IN APPLIED BIOLOGY (AB000004)

### 1. language

English

### 2. course contents

Coordinator: Prof. Alessandro ARCOVITO

Year Course: 1° year

Semester: 1° semester

CFU/UFC: 10

Modules and lecturers:

- FUNDAMENTALS OF PROGRAMMING FOR BIOLOGICAL SCIENCES (AB000036) - 3 cfu - ssd ING-INF/05

Prof. Daniele TOTI

- PROTEIN MODELLING (AB000038) - 2 cfu - ssd BIO/10

Prof. Alessandro ARCOVITO

- BASIC OF IMAGING AND SPECTROSCOPY (AB000041) - 3 cfu - ssd FIS/03

Prof. Gabriele FERRINI

- MACHINE LEARNING FOR BIOLOGICAL SCIENCES (AB000043) - 2 cfu - ssd ING-INF/05

Prof. Daniele TOTI

### 3. BIBLIOGRAPHY

For the Basic of imaging and spectroscopy module, a reference book is Murphy, Douglas B., and Michael W. Davidson. *Fundamentals of light microscopy and electronic imaging*. John Wiley & Sons, 2012. The material will be provided by the teacher, consisting mainly in slides and scientific articles.

For the Modules of Fundamentals of Programming for Biological Sciences and Machine Learning for Biological Sciences, the material will be provided by the teacher and will consist of slides and manuals for further study.

For the Protein Modeling module, the material will be provided by the teacher and will consist of scientific articles.

### 4. LEARNING OBJECTIVES

This course aims to provide students a practical and theoretical knowledge of some of the analytical and quantitative tools applied in regenerative medicine to facilitate the understanding and analysis of biological phenomena. The course is focused on the key concepts of bioinformatics applied to the biological field, such as programming (different programming languages, designing and analyzing algorithms) and on the informatic tools used to predict the structure of proteins starting from their amino acid sequence up to obtain detailed information about their function. The course will also introduce the principles of different machine learning techniques and their applications in the biomedical field for the extrapolation of relevant information from large datasets (data mining and artificial intelligence). The course also includes a module focused on the various imaging techniques and mass spectrometry used in biomedical research to study morphology, molecular content, and processes in biological tissues, both in physiological and pathological conditions.

## **Knowledge and understanding**

The course aims to provide the fundamental knowledge of analytical and quantitative tools to formally and computationally describe biological phenomena and provide models for the analysis and interpretation of experimental data. To know the fundamental tasks of bioinformatics such as basic programming, the use of software for the analysis of structures, biological networks, and machine learning methods applicable in regenerative medicine, and to understand the principles underpinning imaging and mass spectrometry.

## **Applying knowledge and understanding**

Understand how programming, the use of software, and the techniques proposed can be employed to analyze and track data in the biomedical field and specifically in translational medicine. Being able to perform simulations in order to apply and implement the techniques acquired to real-world phenomena/datasets and understand how these can be exploited to investigate the therapeutic potential of stem cells.

## **Making judgements**

Know the main analytical tools used in biomedical research and recognize the situations in which they can be used to expand the therapeutic potential of stem cells in regenerative medicine. Evaluate correct reasoning and applications and identify flaws in deductive and experimental processes.

## **Communication skills**

Know how to communicate clearly and unambiguously, to correctly use technical language, to appropriately disseminate and expose both the principles underlying the methods addressed during the course and the results obtained from these techniques. To be able to generate and transmit questions, ideas, and solutions concerning the disciplines addressed, to specialist and non-specialist interlocutors.

## **Learning skills**

The student must be able to demonstrate a good capacity for self-assessment, and to be continuously updated through scientific articles and online platforms (NCBI, ATCC, Human cell atlas, etc.), and scientific meetings.

## **5. prerequisiteS**

Basic knowledge of Physics, Chemistry, and Biochemistry is necessary, as they are integral parts of the undergraduate courses that grant access to this Master's degree program.

## **6. TEACHING METHODS**

The teaching methods used in this course include frontal lectures with the aid of slides and video tutorials, as well as the opportunity to use computer labs for specific practical exercises to acquire principles of programming, software for the analysis of three-dimensional protein structures, and software for image analysis.

- Knowledge and understanding (Dublin 1): The teaching activities conducted with the help of slides and video tutorials will ensure an effective transfer of information.
- Applying knowledge and understanding (Dublin 2): The possibility of practical exercises will allow the assessment of applied knowledge acquisition.
- Making judgments (Dublin 3): Students will be encouraged to work on independent projects within the course, culminating in the completion of written papers that will be evaluated and contribute to the final grade.
- Communication skills (Dublin 4): Students will be invited to present case studies during lectures and showcase projects completed independently or in collaboration with other students to teachers and peers.
- Learning skills (Dublin 5): The opportunity to study autonomously and utilize modern software will ensure comprehensive and up-to-date training for students enrolled in this degree program.

## 7. OTHER INFORMATIONS

Professors receive students by appointment, also using the main remote interview platforms such as Zoom or Teams.

### **NOTE ON STUDENTS' RESPONSIBILITY**

The responsibility for learning falls increasingly on students, as they advance through the course; hence, ultimately, the commitment and the dedication to learn must come from them.

As members of the Università Cattolica S. Cuore learning community, students are expected to respect the intellectual property of course instructors. All course materials presented to students are the copyrighted property of the course instructors and are subject to the following conditions of use:

- 1) Students may not record nor reproduce lectures or any other classroom activities, unless differently specified by the instructor; however, they may use the recordings for their own course-related purposes only.
- 2) Students may not reproduce and/or post any course material provided by the instructors online or distribute them without the advance written permission of the course instructor and, if applicable, of any students whose voice or image is included in the recordings.
- 3) Any students violating the conditions described above may face academic disciplinary sanctions. As members of a learning community, students are expected to respect the time and efforts of their fellow classmates. Therefore, the use of social media and other electronic distractions that can disrupt the concentration of other students in the classroom is NOT allowed.

### **NOTE ON ACADEMIC INTEGRITY AND CHEATING POLICY**

The principles of truth and honesty are fundamental to the educational process and the academic integrity of the University. All students have a right to expect fair and honest evaluation of their work. **CHEATING UNDERMINES THIS EXPECTATION AND WILL NOT BE TOLERATED.**

Students must avoid the following misconduct behaviors that are considered as cheating:

DO NOT exchange ID badges to collect presence among classmates who cannot attend a lecture.

DO NOT share answers of quizzes during exams.

Any student found by the instructors to be cheating will receive a failing grade for the exam or other graded work, and will be reported to the Course's Coordinator and Instructors' Committee. The instructors may, at their discretion, decide to give a failing grade for the course in severe cases of academic dishonesty.

## 8. learning verification methods

The assessment of learning will take place according to the following procedure:

1. First, the student must submit a written paper agreed upon with the teacher of the "Fundamentals of Programming for Biological Sciences" and "Machine Learning for Biological Sciences" modules at least 1 week before taking the exam. The paper will be a software project to be developed in the Python language.
2. Similarly, the student must also submit a written paper to the teacher responsible for the "Basic of Imaging and Spectroscopy Sciences" module at least 1 week before taking the exam.
3. On the day of the exam, the student will take a written test consisting of a multiple-choice quiz of 30 questions (3 questions for each credit unit, with 5 possible answers and only 1 correct answer), to be completed in 30 minutes. To pass the written test, the student must answer at least 18 out of 30 questions correctly.

The final grade will be determined by the score of the quiz expressed on a scale of thirty, to which an evaluation of the previously submitted written papers will be added. Each written

paper will contribute with a score of + or - 2 thirty. Therefore, the FINAL GRADE will be the GRADE OF THE WRITTEN TEST + or - 4 thirty. The minimum passing grade is 18/30, and the maximum grade is 30/30 with honors, which is achieved by obtaining at least 31/30 by combining the scores of the multiple-choice written test and the previously submitted papers. This examination method fully achieves the objectives of the Berlin descriptors regarding evaluation methods, as follows:

- Knowledge and understanding (Dublin 1): It is achieved by the opportunity to assess the student's ability to write a specific paper on course topics.
- Knowledge and understanding applied (Dublin 2): It is achieved through the ability to answer a multiple-choice test on the entire course program.
- Judgement (Dublin 3): It is achieved through the ability to answer a multiple-choice test on the entire course program.
- Communication skills (Dublin 4): It is achieved by the opportunity to assess the student's ability to write a specific paper on course topics.
- Learning skills (Dublin 5): It is achieved by combining different assessment methods into a single aggregated grade that measures this learning ability consistently.

## 9. program

### **< FUNDAMENTALS OF PROGRAMMING FOR BIOLOGICAL SCIENCES - AB000036 >**

This course covers the following areas: (1) Basic Python syntax, assignments, variables, operators, and introduction to PyCharm IDE. (2) The definition, usage, and debugging of different types of functions, including Python's built-in and custom libraries. (3) An exploration of data structures such as strings and collections, alongside the concept of 'objects' in Python. (4) Deeper understanding of function concepts, file, and memory management, alongside advanced exception handling. (5) Principles of object-oriented programming including classes, objects, abstraction, encapsulation, inheritance, polymorphism, and their implementation in Python. (6) Advanced object-oriented concepts, design patterns, and software architectures, alongside principles of recursion and date/time management. (7) Introduction to Numpy, Matplotlib, and Pandas libraries for scientific computation and data visualization. (8) Introduction to biomedical-focused libraries like RDKit and BioPython, including an overview of free biomedical research software.

### **< MACHINE LEARNING FOR BIOLOGICAL SCIENCES - AB000043 >**

This course covers the following areas: (1) Introduction to machine learning concepts and types, along with the workflow. (2) Data preprocessing techniques including handling missing data, outliers, normalization, standardization, data splitting, and cross-validation. (3) Supervised learning algorithms, with practical applications on biological datasets. (4) Unsupervised learning methods and their applications in the biomedical field. (5) Fundamentals of neural networks and deep learning, focusing on their use for drug discovery and protein function prediction.

### **< BASIC OF IMAGING AND SPECTROSCOPY - AB000041 >**

Image formation and contrast generation in an optical microscope: light, lenses and geometrical optics, Abbe's theory, diffraction and the Airy disk, the point spread function, the spatial resolution. Phase contrast and darkfield microscopy. Polarization microscopy, Differential Interference Contrast (DIC) and modulation contrast microscopy. Fluorescence Microscopy (FM): physical basis of fluorescence, filters and the epi-illumination in FM, objectives and spatial resolution in FM, quenching, blinking, and photobleaching. Detection and image analysis: fundamentals of digital imaging, digital image processing. Specialized microscopies: confocal laser scanning microscopy, two-photon microscopy, superresolution imaging. Brief introduction to Transmission Electron Microscopy (TEM), Cryocrystallography, Mass Spectrometry (application: Mass Cytometry).

### **< PROTEIN MODELLING - AB000038 >**

Theoretical models, introduction. Molecular models in 3D graphics: methodologies for representing molecules, atoms, and their properties; representations of molecular surfaces and volumes. Databases of three-dimensional and two-dimensional structures. Approaches to building molecular models: quantum approaches (overview, advantages, and limitations) and approaches based on classical physics. Force fields in describing molecular properties (all-atoms, united-atoms, and coarse-grained force fields). Methods for calculating the structure and properties of small molecules and proteins. Molecular docking: theoretical study of interactions between molecules; protein-ligand and protein-protein docking; most common algorithms in solving the docking problem. Molecular dynamics.