



# UNIVERSITÀ DEGLI STUDI DI PALERMO

|                              |   |
|------------------------------|---|
| DEPARTMENT                   | Matematica e Informatica  |
| ACADEMIC YEAR                | 2017/2018   |
| MASTER'S DEGREE (MSC)        | COMPUTER SCIENCE  |
| SUBJECT                      | ALGORITHM SCIENCE AND ENGINEERING   |
| TYPE OF EDUCATIONAL ACTIVITY | B   |
| AMBIT                        | 50341-Discipline Informatiche   |
| CODE                         | 06321   |
| SCIENTIFIC SECTOR(S)         | INF/01  |
| HEAD PROFESSOR(S)            | GIANCARLO RAFFAELE Professore Ordinario Univ. di PALERMO  |
| OTHER PROFESSOR(S)           |   |
| CREDITS                      | 6   |
| INDIVIDUAL STUDY (Hrs)       | 102   |
| COURSE ACTIVITY (Hrs)        | 48  |
| PROPAEDEUTICAL SUBJECTS      |   |
| MUTUALIZATION                |   |
| YEAR                         | 1   |
| TERM (SEMESTER)              | 2° semester   |
| ATTENDANCE                   | Not mandatory   |
| EVALUATION                   | Out of 30   |
| TEACHER OFFICE HOURS         | <b>GIANCARLO RAFFAELE</b><br>Monday 15:00 17:00 Stanza 106 Dipartimento di Matematica ed Informatica<br>Thursday 15:00 17:00 Stanza 106 Dipartimento di Matematica ed Informatica |

**DOCENTE:** Prof. RAFFAELE GIANCARLO

|                               |  |
|-------------------------------|--|
| <b>PREREQUISITES</b>          | At least one course of programming and of algorithms and data structures. Moreover, basic knowledge of algebra and geometry are required.  |
| <b>LEARNING OUTCOMES</b>      | <p>Knowledge and Understanding</p> <p>Acquisition of advanced methods for the design and analysis of algorithms. Ability to use the specific language proper of this specialised area.</p> <p>Ability to apply knowledge and understanding.<br/>Ability to develop software based on efficient algorithms for large datasets.</p> <p>Autonomy of Judgment<br/>To be able to evaluate the implications and the results of algorithmic studies and of computational complexity associated to those topics..</p> <p>Communication Abilities<br/>Ability to explain algorithmic results even to a non-expert audience. To be able to illustrate the technological fall-out coming from algorithmic theory.</p> <p>Ability to Learn<br/>Ability to follow-up the advances of the field via the consultation of advanced texts and publications proper of the subject area.</p>  |
| <b>ASSESSMENT METHODS</b>     | <p>Oral Exam</p> <p>The oral exam consists of a discussion on the algorithmic techniques presented during the lectures, on the mathematical laws connected to them. Its goal is to verify, in addition to the knowledge acquired by the candidate, the degree of exposition of the material in the course.<br/>The final grade, suitably scaled as follows, will be based on the following conditions:</p> <p>a) Basic knowledge of advanced algorithmics for management of BIG DATA, and limited ability to apply those techniques autonomously in new contexts, sufficient ability of exposition of the techniques and concepts associated to them (grade (18-21);</p> <p>b) Good knowledge of advanced algorithmics for management of BIG DATA, and good ability to apply those techniques autonomously in new contexts, good ability of exposition of the techniques and concepts associated to them (grade (22-25);</p> <p>c) Deep knowledge of advanced algorithmics for management of BIG DATA, and very good ability to apply those techniques autonomously in new contexts, very good ability of exposition of the techniques and concepts associated to them (grade (26-28);</p> <p>d) Excellent knowledge of advanced algorithmics for management of BIG DATA, and excellent ability to apply those techniques autonomously in new contexts, excellent ability of exposition of the techniques and concepts associated to them (grade (29-30L);</p> |
| <b>EDUCATIONAL OBJECTIVES</b> | To expose the student to the advanced techniques for the design and analysis of computer algorithms. In particular, the entire spectrum of dynamic data structures is covered, with an in-depth study of the intrinsic computational complexity of difficult problems or problems that involve a large quantity of data.   |
| <b>TEACHING METHODS</b>       | Class Lectures in which advanced tools for the analysis of algorithms are provided, being followed by the discussion of problems for the management of Big Data.   |
| <b>SUGGESTED BIBLIOGRAPHY</b> | <p>William J. Cook, William H. Cunningham, William R. Pulleyblank, Alexander Schrijver. Combinatorial Optimization, Wiley 1997<br/>Per le parti che riguardano algoritmi di approssimazione- For the part regarding approximation algorithms</p> <p>Robert Endre Tarjan. Data Structure and Network Algorithms, SIAM 1984<br/>Per le parti che riguardano strutture dati in memoria interna-for the part regarding data structures in internal memory</p> <p>Camil Demetrescu, Irene Finocchi, Giuseppe F. Italiano, Algoritmi e Strutture dati, McGraw Hill, 2005<br/>per le parti che riguardano l'analisi ammortizzata di algoritmi. For the part regarding amortised analysis of algorithms.</p> <p>H. Cormen. C. Leiserson, R. Rivest, C. Stein Introduzione agli algoritmi e strutture dati, McGraw Hill, 2001</p>   |

|  |   |
|--|---|
|  | <p>Per le parti che riguardano hashing e schemi di approssimazione-for the parts regarding hashing and approximation schemes</p> <p>Materiale distribuito dal docente. Tutto il resto del syllabus. The remaining part of the syllabus.</p> |
|--|---|

## SYLLABUS

| Hrs | Frontal teaching   |
|-----|--|
| 3   | <p>Preliminary Notions and Background</p> <p>Amortised Analysis of Algorithms: the method of credits; the potential method. Experimental analysis of Algorithms.</p>   |
| 6   | <p>Internal Memory Advanced Data Structures: Balancing</p> <p>Red-Black Trees and analysis of the operations they support. Linking and Cutting of Trees.</p>   |
| 6   | <p>ADVANCED DATA STRUCTURES IN INTERNAL MEMORY: Amortization</p> <p>Self-adjusting binary trees and analysis of the operations they support.</p> <p>Self-organizing Data Structure.<br/>Self-organizing List.</p>  |
| 4   | <p>Space Succinct Dictionaries</p> <p>Universal Hashing. Bloom Filters</p>   |
| 5   | <p>External Memory Algorithms.</p> <p>Motivation. External Memory Computational Model and the Memory Hierarchy. Sorting and Searching for large datasets.</p>  |
| 6   | <p>Steaming Model</p> <p>Motivation. The straming model of computation. Examples of streaming algorithms for mining large datasets. Synoptic Data Structures</p>   |
| 6   | <p>Data Compression Schemes and their analysis.</p> <p>Static and adaptive data compression schemes.<br/>Compression Boosting and its engineering. Efficient data structures for data compression. Benchmarks for data compression performance analysis.</p>                   |
| 6   | <p>NP-Completness Theory and Polynomial Approximations.</p> <p>Polynomial time approximation schemes. Non-approximability. The classes P, NP and Max-SNP Hard.</p>   |
| 6   | <p>Approximation Algorithms and Heuristic Methods.</p> <p>TSP with triangle inequality. Nearest Neighbour method, Insertion Method, Christofdes heuristics. Tour improvement algorithms: 2-opt, 3-opt, Lin-Kernaghan method, Chaiin Lin-Kernaghan. Held- Karp lower bound.</p> |