

Content

| Course Code | Course Name | Semester | Theory | Practice | Lab | Credit | ECTS |
|-------------|--------------|----------|--------|----------|-----|--------|------|
| VM 521 | Optimisation | 1 | 4 | 0 | 0 | 3 | 8 |

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| Prerequisites | |
| Admission Requirements | |

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| Language of Instruction | English |
| Course Type | Compulsory |
| Course Level | Masters Degree |
| Objective | Our first aim in this course is to learn the mathematical construction and solution methods of optimization problems under constraints or without constraints. Secondly, it is to address the optimization problems encountered in Data Science as an application. |
| Content | <p>Introduction to Mathematical Definitions and Concepts</p> <p>Convexity</p> <p>Derivative</p> <p>Taylor polynomials</p> <p>Unconstrained Optimization</p> <p>Local vs global problem</p> <p>Primary and secondary conditions</p> <p>Algorithms, two basic strategies: line search and trust region</p> <p>Least Squares Problems-Regression Application</p> <p>Optimization Under Constraints</p> <p>feasible region</p> <p>Equality constraint-Inequality constraint and Lagrange method</p> <p>Geometric View</p> <p>Linear programming-Quadratic Programming</p> <p>Simplex method, dual problem</p> <p>Interior points method</p> <p>Application: Machine Learning Problems</p> <p>Clustering-Binary classification-Audio processing-Recommendation Systems-Logistic correlation-Deep learning-Artificial neural networks..etc.</p> |

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| References | <p>Numerical Optimization, J. Nocedal& S. J. Wright, Springer, 1999. ve 2. basım: 2006.</p> <p>Introduction to Global Optimization, R. Horst , P. M.Pardolas&N. V. Thoai, Kluwer Academic Publishers, 1995.</p> <p>The Princeton Companion to Applied Mathematics, Edited by Nicholas J. Higham, Princeton University Press, 2015</p> <p>https://nhigham.com/2016/03/29/the-top-10-algorithms-in-applied-mathematics/</p> <p>Linear Programming and Network Flows, Mokhtar S. Bazaraa, John J. Jarvis, Hanif D. Sherali. John Wiley, 2004. Third edition</p> <p>A gentle introduction to optimization / B. Guenin , J. Könemann , L. Tunçel Cambridge University Press</p> <p>http://www.veridefteri.com/: en güncel kaynaklar, ders notları, haber, bilimsel programlama</p> |
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Theory Topics

| Week | Weekly Contents |
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| 1 | Introduction to the course syllabus and the relationship between Data Science and Optimization |
| 2 | Introduction to Mathematical Definitions and Concepts, Convexity. Derivative. Taylor polynomials. |
| 3 | Unrestricted Optimization. Local vs global problem. Primary and secondary conditions. Problem Application. |
| 4 | Numerical Methods and Algorithms. Least Squares Problems-Regression Application. |
| 5 | Optimization Under Constraints, Feasible region, Equality-Inequality constraints. Lagrange multiplier method. |
| 6 | Geometric View and Applications |
| 7 | Midterm |
| 8 | Linear programming. Simplex method, dual problem |
| 9 | Quadratic Programming. Problems. |
| 10 | Application: Artificial Learning Problems Clustering-Binary classification-Audio processing-Recommendation Systems- Logistic correlation-Deep learning-Artificial neural networks..etc |
| 11 | Application: Artificial Learning Problems Clustering-Binary classification-Audio processing-Recommendation Systems- Logistic correlation-Deep learning-Artificial neural networks..etc |