



## Course Syllabus: Numerical Optimization - AMCS 211

|  |  |
|--|--|
| <b>Division</b>                        | Computer, Electrical and Mathematical Sciences & Engineering |
| <b>Course Number</b>                   | AMCS 211   |
| <b>Course Title</b>                    | Numerical Optimization                                       |
| <b>Academic Semester</b>               | Spring   |
| <b>Academic Year</b>                   | 2017/2018  |
| <b>Semester Start Date</b>             | 01/28/2018   |
| <b>Semester End Date</b>               | 05/24/2018   |
| <b>Class Schedule</b><br>(Days & Time) | 01:00 PM - 02:30 PM   Mon Wed                                |

| Instructor(s)         |                             |               |                                 |              |
|-----------------------|-----------------------------|---------------|---------------------------------|--------------|
| Name                  | Email                       | Phone         | Office Location                 | Office Hours |
| Bernard Semaan Ghanem | Bernard.Ghanem@kaust.edu.sa | +966128080261 | 2125, 1, Al-Khwarizmi (bldg. 1) | TBD          |

| Teaching Assistant(s) |       |
|-----------------------|-------|
| Name                  | Email |
| TBD                   | TBD   |

| Course Information                           |  |
|--|--|
| <b>Comprehensive Course Description</b>      | This course studies fundamental concepts of optimization from two viewpoints: theory and algorithms. It will cover ways to formulate optimization problems (e.g. in the primal and dual domains), study feasibility, assess optimality conditions for unconstrained and constrained optimization, and describe convergence. Moreover, it will cover numerical methods for analyzing and solving linear programs (e.g. simplex), general smooth unconstrained problems (e.g. first-order and second-order methods), quadratic programs (e.g. linear least squares), general smooth constrained problems (e.g. interior-point methods), as well as, a family of non-smooth problems (e.g. ADMM). |
| <b>Course Description from Program Guide</b> | Solution of nonlinear equations. Optimality conditions for smooth optimization problems. Theory and algorithms to solve unconstrained optimization; linear programming; quadratic programming; global optimization; general linearly and nonlinearly constrained optimization problems.  |

|                                  |  |
|----------------------------------|--|
| <b>Goals and Objectives</b>      | <p>At the end of this course, students should:</p> <ul style="list-style-type: none"> <li>-be able to formulate problems in their fields of research as optimization problems by defining the underlying independent variables, the proper cost function, and the governing constraint functions.</li> <li>-be able to transform an optimization problem into its standard form as outlined in the course.</li> <li>-understand how to assess and check the feasibility and optimality of a particular solution to a general constrained optimization problem.</li> <li>-be able to evaluate whether the cost function and the constraints are convex, thus defining a convex problem with strong guarantees on optimality and convergence.</li> <li>-be able to use the optimality conditions to search for a local or global solution from a starting point.</li> <li>-be able to formulate the dual problem of some general optimization types and assess their duality gap using concepts of strong and weak duality.</li> <li>-understand the computational details behind the numerical methods discussed in class, when they apply, and what their convergence rates are.</li> <li>-be able to implement the numerical methods discussed in class and verify their theoretical properties in practice.</li> <li>-be able to apply the learned techniques and analysis tools to problems arising in their own research.</li> </ul>                                       |
| <b>Required Knowledge</b>        | Prerequisites include multivariate calculus, elementary real analysis, and linear algebra.   |
| <b>Reference Texts</b>           | <p>Required Textbook:</p> <ul style="list-style-type: none"> <li>-<i>Numerical Optimization</i>, J. Nocedal and S. Wright, Springer Series in Operations Research and Financial Engineering, 2006</li> </ul> <p>Reference Books:</p> <ul style="list-style-type: none"> <li>-<i>Linear Programming with MATLAB</i>, M. Ferris, O. Mangasarian, and S. Wright, MPS-SIAM Series on Optimization, 2007</li> <li>-<i>Convex Optimization</i>, S. Boyd and L. Vandenberghe, Cambridge University Press, 2004</li> </ul>   |
| <b>Method of evaluation</b>      | <p>5.00% - Course Project(s)<br/> 30.00% - Final exam<br/> 30.00% - Homework /Assignments<br/> 30.00% - Midterm exam<br/> 5.00% - Quiz(zes)</p>  |
| <b>Nature of the assignments</b> | <p><b>Homework and Quizzes</b></p> <p>There will be homework assignments every two weeks, which include programming problems. The handed-in assignment will be corrected in a timely manner and solutions will be provided by the instructor thereafter. Drop quizzes will be administered at the beginning of some classes at the discretion of the instructor to make sure the students are following the course material. Therefore, student attendance and pre-class preparation is very important. It is expected that each student does his/her own assignment individually. Copying homeworks is not tolerated and will be dealt with accordingly.</p> <p><b>Exams</b></p> <p>Two exams are scheduled during the semester, outside of class hours. The date and time of the midterm exam will be agreed upon via a unanimous vote. The exams are closed book, but each student is allowed one A4 hand-written "cheat sheet" for the midterm exam and two such sheets for the final. The content of these sheets is at the discretion of the student.</p> <p><b>Project</b></p> <p>The end of semester project gives each student to opportunity to apply the concepts and methods taught in class to optimization problems they encounter in their own research. Each student will propose their own project, upon the consent of the instructor. If a student cannot come up with a feasible topic for their project, the instructor will propose one for him/her.</p> |
| <b>Course Policies</b>           | All homework assignments, quizzes, and exams are required. Students who do not show up for a quiz or an exam should expect a grade of zero. If you dispute your grade on any homework, quiz, or exam, you may request a re-grade (from the TA for the homeworks and quizzes or from the instructor for the exams) only within 48 hours of receiving the graded exam. Incomplete (I) grade for the course will only be given under extraordinary circumstances such as sickness, and these extraordinary circumstances must be verifiable. The assignment of an (I) requires first an approval of the dean and then a written agreement between the instructor and student specifying the time and manner in which the student will complete the course requirements.   |

|                               |  |
|-------------------------------|--|
| <b>Additional Information</b> | <p>Optimization is at the core of many fields in applied mathematics, engineering, and computer science. For example, engineers want to design the “best” system that has a certain desirable behavior, while remaining faithful to the design specifications. This inherently describes an optimization problem. Once formulated and modeled, knowledge of feasibility, optimality, and numerical methods to achieve both is needed. As such, this course teaches students the building blocks to find the “best” solutions they are seeking.</p> <p>Although this course highlights fundamental points that are needed for a deeper study of the field of optimization, it obviously cannot cover all aspects of this topic. Therefore, it is the student’s responsibility to take initiative and pursue external readings and exercises (self-study) to better understand the rich material being conveyed and to appreciate its impact on the research process more.</p> |
|-------------------------------|--|

### Tentative Course Schedule

*(Time, topic/emphasis & resources)*

| Week | Lectures                         | Topic   |
|------|----------------------------------|---|
| 1    | Mon 01/29/2018<br>Wed 01/31/2018 | Background review   |
| 2    | Mon 02/05/2018<br>Wed 02/07/2018 | Background review<br>Introduction to optimization (general formulations, optimality, and feasibility) |
| 3    | Mon 02/12/2018<br>Wed 02/14/2018 | Unconstrained optimization theory (optimality conditions)   |
| 4    | Mon 02/19/2018<br>Wed 02/21/2018 | Convex functions and convex sets  |
| 5    | Mon 02/26/2018<br>Wed 02/28/2018 | Line search optimization  |
| 6    | Mon 03/05/2018<br>Wed 03/07/2018 | Trust region optimization   |
| 7    | Mon 03/12/2018<br>Wed 03/14/2018 | Popular methods: gradient and conjugate gradient descent  |
| 8    | Mon 03/19/2018<br>Wed 03/21/2018 | Popular methods: Newton and quasi-Newton methods  |
| 9    | Mon 03/26/2018<br>Wed 03/28/2018 | Least squares optimization (linear and nonlinear)   |
| 10   | Mon 04/02/2018<br>Wed 04/04/2018 | Constrained optimization theory (optimality conditions)   |
| 11   | Mon 04/09/2018<br>Wed 04/11/2018 | Lagrangian formulation, duality, and dual problems  |
| 12   | Mon 04/16/2018<br>Wed 04/18/2018 | Interior point methods  |
| 13   | Mon 04/23/2018<br>Wed 04/25/2018 | Linear programming  |
| 14   | Mon 04/30/2018<br>Wed 05/02/2018 | Linear programming  |
| 15   | Mon 05/07/2018<br>Wed 05/09/2018 | Quadratic programming   |
| 16   | Mon 05/14/2018<br>Wed 05/16/2018 | Augmented Lagrangian methods  |
| 17   | Mon 05/21/2018<br>Wed 05/23/2018 | Augmented Lagrangian methods<br>Majorization Minimization   |
| 18   |                                  | Review  |

**Note**

The instructor reserves the right to make changes to this syllabus as necessary.