

TMA946 Applied optimization TM, 5 credits MAN280 Optimization, 5 credits

The course is a basic course in optimization.

The purpose of the course is to provide

- (I) knowledge of some important classes of optimization problems and of application areas of optimization modelling and methods;
- (II) practice in describing relevant parts of a real-world problem in a mathematical model;
- (III) understanding the basic mathematical theory upon which optimality criteria are built;
- (IV) examples of optimization algorithms that are naturally developed from this theory, their analysis, and their application to practical optimization problems.

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Course presentation

CONTENTS: The course is broad in its contents, but has its main focus on optimization problems in continuous variables, and can within this framework be separated into two main areas:

Linear programming: Linear optimization models, linear programming theory and geometry, the Simplex method, duality, interior point methods, sensitivity analysis, modelling languages;

Nonlinear programming: Nonlinear optimization models, convexity theory, optimality conditions, iterative algorithms for optimization problems with or without constraints, relaxations, penalty functions, Lagrangian duality.

An overview is also given of three other important problem areas within optimization: integer programming, network optimization, and optimization under uncertainty.

ORGANIZATION: Lectures, exercises, computer exercises, and a project assignment.

COURSE LITERATURE:

- (i) Linear and Nonlinear Programming (book; S G Nash and A Sofer, 1996) [Cremona] Errata: <http://www.gmu.edu/departments/ore/books/naso/ns-errata.html>.
- (ii) Hand-outs from books
- (iii) Collection of exercises [sold by the department]

COURSE REQUIREMENTS: The course content is defined by the literature references in the plan below.

EXAMINATION:

- Written exam (first opportunity 10/3, morning., V house, additional exams in May and August)
- Project assignment
- Två correctly solved computer exercises

SCHEDULE:

Lectures: Normally on Tuesdays 13.15–15.00 and Thursdays 8.00–9.45 in the lecture hall, MD house. *Exceptions: Lectures 1 and 2 are on 21/1 13.15–17.00; no lecture 11/2 (Charm).* Lectures are given in English.

Exercises: In two parallel groups; teaching in Swedish (Niclas/Fredrik) normally Tuesdays 15.15–17.00 and Thursdays 10.00–11.45 in MD6, exercises in English (Anton) normally Tuesdays 15.15–17.00 and Thursdays 10.00–11.45 in MD9. *Exceptions: no exercise 21/1 (see above); no exercise 11/2 (Charm).*

Project: Teachers are available for questions in the computer rooms, which are also booked for work on the project, on 17/2 (rooms: B, C, MD2), at 17.15–21.00. (Presence is not obligatory.) At other times, work is done individually. Deadline for handing in the project model: 30/1. Hand-out of correct AMPL model: 7/2. Deadline for handing in the project report: 21/2.

Computer exercises: Each computer exercise is scheduled to take place in the computer rooms when also teachers are available, on 10/2 and 27/2, respectively (rooms booked: B, C, MD2), and on both occasions at 17.15–21.00 (Room MD2 is booked from 19.00). (Presence is not obligatory.) The computer exercises can be done individually, but preferably in groups of two. Deadline for handing in the report: one week following each computer exercise.

Information about the project assignment and the computer exercises are found on the web page for the course, <http://www.md.chalmers.se/~mipat/T0kurs.html>.

This course information and other hand-outs will also be found on this page.

COURSE PLAN, LECTURES:

Le 1 (21/1) *Course presentation.* Subject description. Applications. Modelling. Problem analysis. Classification.

(i): Chapter 1

(ii): Course information (this document); course notes, including a document with additional applications

Le 2 (21/1) *Linear programming.* Solving a simple linear program with DUPLO. *Fundamentals.* Local and global optimality. Convex sets and functions.

(i): Chapter 4.1–2, 2.2–3

(ii): “Introduction to linear programming”

Le 3 (23/1) *Linear programming.* Problem description. Polyhedral sets. Basic solutions and extreme points (algebra versus geometry in linear programming). Representation theorem. The Simplex method, introduction.

(i): Chapter 4.3–4, 5.1–2

Le 4 (28/1) *Linear programming.* The Simplex method. Matrix and tableau forms. Phase 1 and 2. Degeneration. Termination. Revised simplex. Complexity.

(i): Chapter 5.3–5.6, 9.3

Le 5 (30/1) *Linear programming.* Optimality. Duality. Sensitivity analysis.

(i): Chapter 6.1–4, 7.4–7.6

Le 6 (4/2) *Linear programming.* Modelling languages and LP solvers.

Integer programming. Applications. Modelling. Heuristics.

(ii): On integer programming

Le 7 (6/2) *Unconstrained optimization.* Optimality. Search methods. Search directions. Line searches. Termination criteria. Steepest descent.

(i): Chapter 10.1–5, 11.4.1, 11.5, 11.1–2

Le 8 (13/2) *Unconstrained optimization.* Newton methods. Derivative-free methods for unconstrained optimization.

Constrained optimization. Primal optimality conditions.

(i): Chapter 11.3, 12.2–3, 12.6, 11.4.3, 14.1

(ii): Material on derivative-free optimization

Le 9 (18/2) *Constrained optimization.* Primal algorithms. Gradient projection. The Frank–Wolfe method. Simplicial decomposition. Gradient projection. Dual optimality conditions (KKT), introduction.

(i): Chapter 14.2–7

(ii): On primal algorithms

Le 10 (20/2) *Constrained optimization.* Farkas’ Lemma. Fritz Johns’ conditions. The Karush–Kuhn–Tucker conditions. Lagrangian duality, introduction.

(i): Chapter 14.2–7, 14.8.2–3

Le 11 (25/2) *Constrained optimization.* Lagrangian duality. Weak and strong duality. Duality gaps. Nondifferentiable optimization.

(i): Chapter 14.8.2–3

(ii): On algorithms for Lagrangian dual problems

Le 12 (27/2) *Constrained optimization.* Penalty and barrier methods. Interior point methods for linear programming, orientation. Generalized reduced gradient (GRG), orientation. The augmented Lagrangian.

(i): Chapter 16.1–3, 17.3–4, 16.5, 15.5, 16.6

(ii): On algorithms for constrained optimization

Le 13 (6/2) *Constrained optimization.* Algorithms for constrained optimization, continued.

Le 14 (6/3) *An overview of the course.*

An orientation on integer and network optimization.

(i): Chapter 8.1–5

(ii): On applications and algorithms for network optimization.

COURSE PLAN, EXERCISES:

The below list contains more exercises than will be covered during the exercises. Those that are left out are recommended home work exercises.

Ex 1 (23/1) Modelling. Geometric solution of linear optimization problems.

(i): 4.1.1–2

(iii): 2.1.1–2.1.13, 2.4.5–2.4.7, 3.1.1–3.1.8

Ex 2 (28/1) Convexity.

(i): 2.2.3, 4, 6, 8, 9; 2.3.1–6, 8–17

(iii): 1.1.8, 1.1.9, 1.1.15, 1.1.16, 1.1.20, 1.1.22, 1.1.27–1.1.37

Ex 3 (30/1) The Simplex method. Geometry, bases.

(i): 4.2.1–6; 4.3.1–3, 6, 10; 4.4.6–8

(iii): 2.3.2, 2.3.11a, 2.3.16, 2.4.4, 2.4.13a, 2.4.19

Ex 4 (4/2) The Simplex method. Matrix form. Phase 1 och 2.

(i): 5.2.2–6, 8; 5.3.4–5; 5.4.2, 4; 5.5.1, 2(a), 3(a), 4, 6–7, 10

(iii): 2.3.1, 2.3.10, 2.3.24, 2.3.25, 2.3.34–36, 2.4.1–2.4.3, 2.3.5–2.3.9.

Ex 5 (6/2) The Simplex method. Optimality. Duality. Sensitivity analysis.

(i): 6.1.1–7; 6.2.1–5, 7, 9, 11–12, 15; 5.2.8; 6.3.1–2, 4–9; 6.4.1–3; 7.5.3–5

(iii): 2.2.1–2.2.5, 2.3.11–2.3.14, 2.2.6–2.2.9, 2.4.9, 2.3.33

Ex 6 (13/2) Duality, continued. Integer programming. Modelling. Heuristics.

(iii): 5.1.4, 11, 15, 17, 5.2.7

Ex 7 (18/2) Unconstrained optimization.

(i): 10.2.1–7, 10–12, 17–18; 10.5.2–6, 9; 11.4.7–9; 11.1.1–2, 6–7, 10; 10.3.2–4, 6, 12

(iii): 3.3.1–3.3.11, 3.2.1

Ex 9 (20/2) Unconstrained optimization, continued.

Ex 10 (25/2) Optimality conditions.

(i): 14.2.4–6; 14.4.4; 14.5.3

(iii): 3.5.1–3.5.7, 3.5.17–3.5.19, 3.5.24

Ex 11 (27/2) Lagrangian duality.

(i): 14.8.4–8, 10

(iii): 3.4.1, 3.5.8

Ex 6 (13/2) Lagrangian duality, continued. Penalty and barrier methods.

(i): 16.2.1–4; 16.5.1–2; 15.5.1–3, 5; 16.6.1, 5–6

Ex 12 (7/3) Penalty and barrier methods. Repetition.

(iii): 6.1.1–6.1.6 and exam questions