DM545 Linear and Integer Programming

> Lecture 1 Introduction

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Department of Mathematics & Computer Science University of Southern Denmark Outline

Course Introduction Introduction Solving LP Problems

1. Course Introduction

2. Introduction

Resource Allocation Diet Problem Duality

3. Solving LP Problems Fourier-Motzkin method

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Context

Course Introduction Introduction Solving LP Problems

Students:

- Computer Science (3rd year)
- Applied Mathematics (3rd year)
- Math-economy (3rd year)

Prerequisites

- Calculus (MM501, MM502)
- Linear Algebra (MM505)

Practical Information

Teacher: Marco Chiarandini (marco@imada.sdu.dk) Instructor: Christian Nørskov (cnoer10@student.sdu.dk)

Schedule (\approx 24 lecture hours + \approx 20 exercise hours):

Week	15	16 17	18	19	20	21	22	23
Tir, 12-14	Intro (Fælles) (U150)		Intro (Fælles) (U150)					
Tir, 14-16							Træning (S1) (U20)	
Tir, 15-17		Intro (Fælles) (U47)						
Ons, 10-12					Intro (Fælles) (U91)			
Ons, 12-14	Intro (Fælles) (U71)	Lab (S1) (Terminalrum)	Træning (S1) (U56)	Intro (Fælles) (U20)	Træning (S1) (U20)	Træning (S1) (U20)	Træning (S1) (U20)	
Fre, 10-12	Træning (S1) (U42)	Intro (Fælles) (U42)		Træning (S1) (U42)		Intro (Fælles) (U42)	Intro (Fælles) (U42)	Træning (S1) (U42)

Communication tools

- ▶ BlackBoard (BB) ⇔ Public Web Page (WWW) (link from http://www.imada.sdu.dk/~marco/DM545/)
- Announcements in BlackBoard
- Classes
- Personal email

Text book

HL Frederick S Hillier and Gerald J Lieberman, Introduction to Operations Research, 9th edition, 2010

Other references:

- MG J. Matousek and B. Gartner. Understanding and Using Linear Programming. Springer Berlin Heidelberg, 2007
- Wo L.A. Wolsey. Integer programming. John Wiley & Sons, New York, USA, 1998
- FGK Robert Fourer, David M. Gay, and Brian W. Kernighan, AMPL: A Modeling Language for Mathematical Programming. Duxbury Press, Brooks Cole, Publishing Company, 2003.

Public Web Page is the main reference for list of contents (pensum). It Contains:

- slides (text missing)
- list of topics
- ▶ links
- software

Contents

Linear Programming

- 1 apr8 Introduction Linear Programming, Notation
- 2 apr9 Linear Programming, Simplex Method
- 3 apr22 Exception Handling
- 4 apr25 Duality Theory
- 5 apr29 Sensitivity
- 6 may6 Revised Simplex Method

Integer Linear Programming

- 7 may7 Modeling Examples, Good Formulations, Relaxations
- 8 may2 Well Solved Problems
- 9 may13 Network Optimization Models (Max Flow, Min cost flow, matching)
- 10 may20 Cutting Planes & Branch and Bound
- 11 may23 More on Modelling
- 12 may27
- 13 may30

Evaluation

► 5 ECTS

- obligatory Assignments, pass/fail, evaluation by teacher (2 hand ins) practical part modeling + programming in AMPL
- 4 hour written exam, 7-grade scale, external censor (theory part) similar to exercises in class on June 10

(language: English and Danish)

- Small projects (in groups of 2) must be passed to attend the written exam
- ► They require the use of the AMPL system + CPLEX or Gurobi Software available for all systems from the Public Web Page: "Software and Data" → "AMPL" (Get the password in class)

Training Sessions

- Prepare them in advance to get out the most
- Best carried out in small groups
- Exam rehearsal (first days of June)

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It seeks to determine how best to design and operate a system, usually under conditions requiring the allocation of scarce resources, by means of **mathematics** and **computer science**.

Quantitative methods for planning and analysis.

Some Examples ...

- Production Planning and Inventory Control
- Budget Investment
- Blending and Refining
- Manpower Planning
 - Crew Rostering (airline crew, rail crew, nurses)
- Packing Problems
 - Knapsack Problem
- Cutting Problems
 - Cutting Stock Problem
- Routing
 - Vehicle Routing Problem (trucks, planes, trains ...)
- Locational Decisions
 - Facility Location
- Scheduling/Timetabling
 - Examination timetabling/ train timetabling
- + many more

Common Characteristics

- Planning decisions must be made
- The problems relate to quantitative issues
 - Fewest number of people
 - Shortest route
- Not all plans are feasible there are constraining rules
 - Limited amount of available resources
- It can be extremely difficult to figure out what to do

OR - The Process?



- 1. Observe the System
- 2. Formulate the Problem
- 3. Formulate Mathematical Model
- 4. Verify Model
- 5. Select Alternative
- 6. Show Results to Company
- 7. Implementation

Central Idea

Build a mathematical model describing exactly what one wants, and what the "rules of the game" are. However, what is a mathematical model and how?

Mathematical Modeling

▶ Find out exactly what the decision makes needs to know:

- which investment?
- which product mix?
- which job j should a person i do?
- Define Decision Variables of suitable type (continuous, integer valued, binary) corresponding to the needs
- Formulate Objective Function computing the benefit/cost
- Formulate mathematical Constraints indicating the interplay between the different variables.

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Resource Allocation

In manufacturing industry, factory planning: find the best product mix.

Example

A factory makes two products standard and deluxe.

A unit of standard gives a profit of 6k Dkk. A unit of deluxe gives a profit of 8k Dkk.

The grinding and polishing times in terms of hours per week for a unit of each type of product are given below:

	Standard	Deluxe
(Machine 1) Grinding	5	10
(Machine 2) Polishing	4	4

Grinding capacity: 60 hours per week Polishing capacity: 40 hours per week **Q**: How much of each product, standard and deluxe, should we produce to maximize the profit?

Mathematical Model

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Decision Variables

 $x_1 \ge 0$ units of product standard $x_2 \ge 0$ units of product deluxe

Object Function

max $6x_1 + 8x_2$ maximize profit

Constraints

 $5x_1 + 10x_2 \le 60$ Grinding capacity $4x_1 + 4x_2 \le 40$ Polishing capacity

Mathematical Model

Machines/Materials A and B Products 1 and 2

Graphical Representation:



a _{ij}	1	2	bi
Α	5	10	60
В	4	4	40
Сј	6	8	

Resource Allocation - General Model

Notation

$$\begin{array}{ll} \max & \sum\limits_{j=1}^{n} c_{j} x_{j} \\ & \sum\limits_{j=1}^{n} a_{ij} x_{j} \leq b_{i}, \ i=1,\ldots,m \\ & x_{j} \geq 0, \ j=1,\ldots,n \end{array}$$

In Matrix Form

$$c^{T} = \begin{bmatrix} c_{1} & c_{2} & \dots & c_{n} \end{bmatrix} \qquad \max \begin{array}{c} z = c^{T} x \\ Ax \leq b \\ x \geq 0 \end{array}$$
$$A = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots \\ a_{31} & a_{32} & \dots & a_{mn} \end{bmatrix}, x = \begin{bmatrix} x_{1} \\ x_{2} \\ \vdots \\ x_{n} \end{bmatrix}, b = \begin{bmatrix} b_{1} \\ b_{2} \\ \vdots \\ b_{m} \end{bmatrix}$$

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Our Numerical Example

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$$\max \sum_{\substack{j=1 \\ j=1}^{n} c_j x_j}^{n} c_j x_j \le b_i, \quad i = 1, \dots, m$$
$$x_j \ge 0, \quad j = 1, \dots, n$$

 $\begin{array}{rll} \max \ c^{T}x \\ Ax \ \leq \ b \\ x \ \geq \ 0 \end{array}$

 $x \in \mathbb{R}^n, c \in \mathbb{R}^n, A \in \mathbb{R}^{m \times n}, b \in \mathbb{R}^m$

max
$$\begin{bmatrix} 6 & 8 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

 $\begin{bmatrix} 5 & 10 \\ 4 & 4 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \leq \begin{bmatrix} 60 \\ 40 \end{bmatrix}$
 $x_1, x_2 \geq 0$

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The Diet Problem (Blending Problems)

- Select a set of foods that will satisfy a set of daily nutritional requirement at minimum cost.
- Motivated in the 1930s and 1940s by US army.
- Formulated as a linear programming problem by George Stigler
- First linear programming problem
- (programming intended as planning not computer code)

min cost/weight subject to nutrition requirements:

> eat enough but not too much of Vitamin A eat enough but not too much of Sodium eat enough but not too much of Calories

. . .



Course Introduction

Solving LP Problems

Suppose there are:

- ▶ 3 foods available, corn, milk, and bread, and
- there are restrictions on the number of calories (between 2000 and 2250) and the amount of Vitamin A (between 5000 and 50,000)

Food	Cost per serving	Vitamin A	Calories
Corn	\$0.18	107	72
2% Milk	\$0.23	500	121
Wheat Bread	\$0.05	0	65

The Mathematical Model

Parameters (given data)

- F = set of foods
- N = set of nutrients
- a_{ij} = amount of nutrient j in food i, $\forall i \in F$, $\forall j \in N$
- c_i = cost per serving of food $i, \forall i \in F$
- F_{mini} = minimum number of required servings of food $i, \forall i \in F$
- F_{maxi} = maximum allowable number of servings of food $i, \forall i \in F$
- N_{minj} = minimum required level of nutrient $j, \forall j \in N$
- N_{maxj} = maximum allowable level of nutrient $j, \forall j \in N$

Decision Variables

 x_i = number of servings of food *i* to purchase/consume, $\forall i \in F$

The Mathematical Model

Objective Function: Minimize the total cost of the food

 $\mathsf{Minimize}\sum_{i\in F}c_ix_i$

Constraint Set 1: For each nutrient $j \in N$, at least meet the minimum required level

$$\sum_{i\in F} a_{ij}x_i \geq N_{minj}, \forall j \in N$$

Constraint Set 2: For each nutrient $j \in N$, do not exceed the maximum allowable level.

$$\sum_{i \in F} a_{ij} x_i \le N_{maxj}, \forall j \in N$$

Constraint Set 3: For each food $i \in F$, select at least the minimum required number of servings

 $x_i \geq F_{mini}, \forall i \in F$

Constraint Set 4: For each food $i \in F$, do not exceed the maximum allowable number of servings.

 $x_i \leq F_{maxi}, \forall i \in F$

The Mathematical Model

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system of equalities and inequalities

$$\begin{array}{ll} \min & \sum_{i \in F} c_i x_i \\ \sum_{i \in F} a_{ij} x_i \geq N_{minj}, & \forall j \in N \\ \sum_{i \in F} a_{ij} x_i \leq N_{maxj}, & \forall j \in N \\ & x_i \geq F_{mini}, & \forall i \in F \\ & x_i \leq F_{maxi}, & \forall i \in F \end{array}$$

- ► The linear programming model consisted of 9 equations in 77 variables
- ► Stigler, guessed an optimal solution using a heuristic method
- In 1947, the National Bureau of Standards used the newly developed simplex method to solve Stigler's model.
 It took 9 clerks using hand-operated desk calculators 120 man days to solve for the optimal solution

The original instance: http://www.gams.com/modlib/libhtml/diet.htm

AMPL Model

AMPL Model

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diet.dat

data;

```
set NUTR := A B1 B2 C ;
set FOOD := BEEF CHK FISH HAM MCH
MTL SPG TUR;
```

```
param: cost f _ min f _ max :=
    BEEF 3.19 0 100
    CHK 2.59 0 100
    FISH 2.29 0 100
    HAM 2.89 0 100
    MCH 1.89 0 100
    MTL 1.99 0 100
    SPG 1.99 0 100
    TUR 2.49 0 100;

param: n min n max :=
```

A 700 10000 C 700 10000 B1 700 10000 B2 700 10000 :

%

```
param amt (tr):

A C B1 B2 :=

BEEF 60 20 10 15

CHK 8 0 20 20

FISH 8 10 15 10

HAM 40 40 35 10

MCH 15 35 15 15

MTL 70 30 15 15

SPG 25 50 25 15

TUR 60 20 15 10 ;
```

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Duality

Resource Valuation problem: Determine the value of the raw materials on hand such that: The company must be willing to sell the raw materials should an outside firm offer to buy them at a price consistent with the market

- z_i value of a unit of raw material i
- $\sum_{i=1}^{m} b_i z_i$ opportunity cost (cost of having instead of selling)
 - ρ_i prevailing unit market value of material *i*
 - σ_j prevailing unit product price

Goal is to minimize the lost opportunity cost

$$\min \sum_{i=1}^{m} b_i z_i$$

$$z_i \ge \rho_i, \quad i = 1 \dots m$$

$$\sum_{i=1}^{m} z_i a_{ij} \ge \sigma_j, \quad j = 1 \dots n$$
(1)
(2)
(3)

(2) and (3) otherwise contradicting market

Let

 $y_i = z_i - \rho_i$

markup that the company would make by reselling the raw material instead of producing.

$$\min \sum_{i=1}^{m} y_i b_i + \sum_{j=1}^{n} c_j x_j$$

$$\sum_{i=1}^{m} y_i a_{ij} \ge c_j, \quad j = 1 \dots n$$

$$y_i \ge 0, \quad i = 1 \dots m$$

$$\max \sum_{j=1}^{n} c_j x_j$$

$$\sum_{j=1}^{n} a_{ij} x_j \le b_i, \quad i = 1, \dots, m$$

$$x_j \ge 0, \quad j = 1, \dots, n$$

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Fourier-Motzkin method

In Matrix Form

$$c^{T} = \begin{bmatrix} c_{1} & c_{2} & \dots & c_{n} \end{bmatrix} \qquad \max \begin{array}{c} z = c^{T} x \\ Ax = b \\ x \ge 0 \end{array}$$

$$A = \begin{bmatrix} a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & & & \\ a_{31} & a_{32} & \dots & a_{mn} \end{bmatrix}, x = \begin{bmatrix} x_2 \\ \vdots \\ x_n \end{bmatrix}, b = \begin{bmatrix} b_2 \\ \vdots \\ b_m \end{bmatrix}$$

Notation

- ▶ N natural numbers, Z integer numbers, Q rational numbers, R real numbers
- ▶ column vector and matrices scalar product: y^Tx = ∑ⁿ_{i=1} y_ix_i
- linear combination

$$\begin{aligned} x \in \mathbb{R}^k \\ x_1, \dots, x_k \in \mathbb{R} \\ \lambda = (\lambda_1, \dots, \lambda_k)^T \in \mathbb{R}^k \end{aligned} \qquad x = \sum_{i=1}^k \lambda_i x_i$$

$$\begin{array}{c} \lambda \geq 0 \\ \lambda^T 1 = 1 \quad \left(\sum_{i=1}^k \lambda_i = 1\right) \\ \lambda \geq 0 \text{ and } \lambda^T 1 = 1 \end{array}$$

conic combination affine combination convex combination

Linear Programming

Abstract mathematical model:

Decision Variables (quantity) eg. x_1 units of 1, x_2 units of 2

Criterion (discriminate among solutions) eg. max profit: $6x_1 + 8x_2$

Constraints (limitations on resources) eg.

 $5x_1 + 10x_2 \le 60$; $4x_1 + 4x_2 \le 40$; $x_1 \ge 0$; $x_2 \ge 0$

objective func.max / min $c^T \cdot x$ $c \in \mathbb{R}^n$ constraints $A \cdot x \geq b$ $A \in \mathbb{R}^{m \times n}, b \in \mathbb{R}^m$ $x \geq 0$ $x \in \mathbb{R}^n, 0 \in \mathbb{R}^n$

Essential features of a Linear program:

- 1. continuity (later, integrality)
- 2. linearity \rightsquigarrow proportionality + additivity
- 3. certainty of parameters

Notions of Computer Science

Algorithm: a finite, well-defined sequence of operations to perform a calculation

Algorithm: LargestNumber



return largest

Running time: proportional to number of operations, eg O(n)

Growth Functions



NP-hard problems: bad if we have to solve them, good for cryptology

History of Linear Programming (LP)

- Origins date back to Newton, Leibnitz, Lagrange, etc.
- In 1827, Fourier described a variable elimination method for systems of linear inequalities, today often called Fourier-Moutzkin elimination (Motzkin, 1937). It can be turned into an LP solver but inefficient.
- In 1932, Leontief (1905-1999) Input-Output model to represent interdependencies between branches of a national economy (1976 Nobel prize)
- In 1939, Kantorovich (1912-1986): Foundations of linear programming (Nobel prize with Koopmans on LP, 1975)
- ► The math subfield of Linear Programming was created by George Dantzig, John von Neumann (Princeton), and Leonid Kantorovich in the 1940s.
- In 1947, Dantzig (1914-2005) invented the (primal) simplex algorithm working for the US Air Force at the Pentagon. (program=plan)

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History of LP (cntd)

- ► In 1954, Lemke: dual simplex algorithm, In 1954, Dantzig and Orchard Hays: revised simplex algorithm
- ► In 1958, Integer Programming was born with cutting planes by Gomory and branch and bound
- In 1970, Victor Klee and George Minty created an example that showed that the classical simplex algorithm has exponential worst-case behaviour.
- In 1979, L. Khachain found a new efficient algorithm for linear programming. It was terribly slow. (Ellipsoid method)
- In 1984, Karmarkar discovered yet another new efficient algorithm for linear programming. It proved to be a strong competitor for the simplex method. (Interior point method)

History of Optimization

- In 1951, Nonlinear Programming began with the Karush-Kuhn-Tucker Conditions
- ▶ In 1952, Commercial Applications and Software began
- ► In 1950s, Network Flow Theory began with the work of Ford and Fulkerson.
- ▶ In 1955, Stochastic Programming began
- ▶ In 1958, Integer Programming began by R. E. Gomory.
- ▶ In 1962, Complementary Pivot Theory

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Fourier Motzkin elimination method

Has $Ax \leq b$ a solution? (Assumption: $A \in \mathbb{Q}^{m \times n}, b \in \mathbb{Q}^n$) Idea:

- transform the system into another by eliminating some variables such that the two systems have the same solutions over the remaining variables.
- 2. reduce to a system of constant inequalities that can be easily decided

Let x_r be the variable to eliminate Let $M = \{1 \dots m\}$ index the constraints For a variable j let partition the rows of the matrix in

 $N = \{i \in M \mid a_{ij} < 0\} \\ Z = \{i \in M \mid a_{ij} = 0\} \\ P = \{i \in M \mid a_{ij} > 0\}$

 $\begin{cases} x_r \ge b_{ir} - \sum_{k=1}^{r-1} a_{ik} x_k \\ x_r \le b_{ir} - \sum_{k=1}^{r-1} a_{ik} x_k \\ \text{all other constraints} (i \in Z) \end{cases}$

$$\begin{cases} x_r \ge A_i(x_1, \dots, x_{r-1}), & i \in N \\ x_r \le B_i(x_1, \dots, x_{r-1}), & i \in P \\ \text{all other constraints}(i \in Z) \end{cases}$$

Hence the original system is equivalent to

 $\begin{cases} \max\{A_i(x_1,\ldots,x_{r-1}), i \in N\} \le x_r \le \min\{B_i(x_1,\ldots,x_{r-1}), i \in P\} \\ \text{all other constraints}(i \in Z) \end{cases}$

which is equivalent to

$$\left(egin{array}{ll} A_i(x_1,\ldots,x_{r-1})\leq B_j(x_1,\ldots,x_{r-1}) & i\in N, j\in P \ all other constraints(i\in Z) \end{array}
ight.$$

we eliminated x_r but:

 $\begin{cases} |N| \cdot |P| \text{ inequalities} \\ |Z| \text{ inequalities} \end{cases}$

after d iterations if |P| = |N| = n/2 exponential growth: $1/4(n/2)^{2^d}$

Example

 x_2 variable to eliminate $N = \{2, 5, 6\}, Z = \{3\}, P = \{1, 4\}$ $|Z \cup (N \times P)| = 7$ constraints Summary

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