

MVE165/MMG631

Linear and integer optimization with applications

Lecture 7

Discrete optimization models and applications;
complexity

Ann-Brith Strömberg

2019-04-09

Recall the diet problem

- Sets

- $\mathcal{J} = \{1, \dots, n\}$ — kinds of food
- $\mathcal{I} = \{1, \dots, m\}$ — kinds of nutrients

- Variables

- $x_j, j \in \mathcal{J}$ — purchased amount of food j per day

- Parameters

- $c_j, j \in \mathcal{J}$ — cost of food j
- $a_j, j \in \mathcal{J}$ — available amount of food j
- $p_{ij}, i \in \mathcal{I}, j \in \mathcal{J}$ — content of nutrient i in food j
- q_i — lower limit on the amount of nutrient i per day
- Q_i — upper limit on the amount of nutrient i per day

The diet problem

The linear optimization model

$$\begin{array}{ll} \text{minimize}_x & \sum_{j=1}^n c_j x_j, \\ \text{subject to} & q_i \leq \sum_{j=1}^n p_{ij} x_j \leq Q_i, \quad i = 1, \dots, m, \\ & 0 \leq x_j \leq a_j, \quad j = 1, \dots, n. \end{array}$$

- What if we may buy at most $k < n$ different kinds of food?
- Define new variables: $y_j = \begin{cases} 1 & \text{if food } j \text{ is in the diet} \\ 0 & \text{otherwise} \end{cases}$
- Model the following relations:

$$y_j = 0 \implies x_j = 0$$

$$y_j = 1 \implies x_j \geq 0$$

The cardinality constrained diet problem

- Add a *cardinality constraint*: $\sum_{j=1}^n y_j \leq k$
- Modify the availability constraints: $0 \leq x_j \leq a_j y_j$

An integer (binary) linear optimization model

$$\begin{array}{ll} \text{minimize}_{x,y} & \sum_{j=1}^n c_j x_j, \\ \text{subject to} & q_i \leq \sum_{j=1}^n p_{ij} x_j \leq Q_i, \quad i = 1, \dots, m, \\ & \sum_{j=1}^n y_j \leq k, \\ & 0 \leq x_j \leq a_j y_j, \quad j = 1, \dots, n, \\ & y_j \in \{0, 1\}, \quad j = 1, \dots, n. \end{array}$$

The cardinality constrained diet problem—an instance

- Buy at most k types of food
- *Totally* $n=20$ types of food:
SourMilk, Milk, Potato, Carrot,
HaricotVerts, GreenBeans,
Spinache, Tomato, Cabbage,
Banana, Queenberries,
OrangeJuice, Chicken, Salmon,
Cod, Rice, Pasta, Egg, Apple,
Ham
- *Constraints on* $m=13$ nutrients:
Energy, Carbohydrates, Fat,
Protein, Fibres, SaturFat,
SingleUnsatFat,
MultiUnsatFat, VitaminD,
VitaminC, Folate, Iron, Salt

Optimal solutions for
 $k \in \{20, 10\}$

k	20	10
Apple	3	3
Banana	2	2
Carrot	2.3	3
Chicken	0.4	--
Egg	2	2
HaricotVerts	0.1	--
Milk	3	3
Pasta	2	2
Potato	2.3	2.4
Rice	1	1
Salmon	0.5	0.8
SourMilk	2	2

*For $k \leq 9$ no feasible
solution exists*

Variables

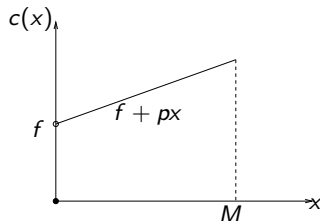
- *Linear programming* (LP) uses continuous variables: $x_{ij} \geq 0$
- *Integer linear programming* (ILP) uses *integer* variables: $x_{ij} \in \mathbb{Z}$
- *Binary linear programming* (BLP) uses *binary* variables: $x_{ij} \in \mathbb{B}$
- If *both* continuous and integer/binary variables are used in a program, it is called a *mixed integer/binary linear program* (MILP)/(MBLP)

Constraints

- An ILP (or MILP) possesses linear constraints and integer requirements on the variables
- Also logical relations, e.g., *if-then* and *either-or*, can be modelled
- This is done by introducing additional (binary) variables and additional constraints

MILP modelling—fixed charges

- Send a truck \Rightarrow Start-up cost: $f > 0$
- Load loafs of bread on the truck \Rightarrow cost per loaf: $p > 0$
- $x = \#$ bread loafs to transport from bakery to store



The cost function $c : \mathbb{R}_+ \mapsto \mathbb{R}_+$ is *nonlinear* and *discontinuous*

$$c(x) := \begin{cases} 0 & \text{if } x = 0 \\ f + px & \text{if } 0 < x \leq M \end{cases}$$

MILP modelling—fixed charges

- Let $y = \#$ trucks to send (here, y equals 0 or 1)
- Replace $c(x)$ by $fy + px$
- Constraints: $0 \leq x \leq My$ and $y \in \{0, 1\}$

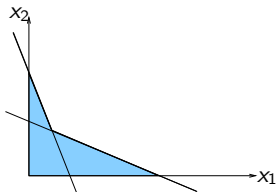
New model:

$$\left[\begin{array}{ll} \min & fy + px \\ \text{s.t.} & x - My \leq 0 \\ & x \geq 0 \\ & y \in \{0, 1\} \end{array} \right]$$

- $y = 0 \Rightarrow x = 0 \Rightarrow fy + px = 0$
- $y = 1 \Rightarrow x \leq M \Rightarrow fy + px = f + px$
- $x > 0 \Rightarrow y = 1 \Rightarrow fy + px = f + px$
- $x = 0 \not\Rightarrow y = 0$ But: Minimization will push y to zero!

Discrete alternatives

- Suppose:
either $x_1 + 2x_2 \leq 4$ *or* $5x_1 + 3x_2 \leq 10$,
and $x_1, x_2 \geq 0$ must hold
- *Not* a convex set



Let $M \gg 1$ and define $y \in \{0, 1\}$

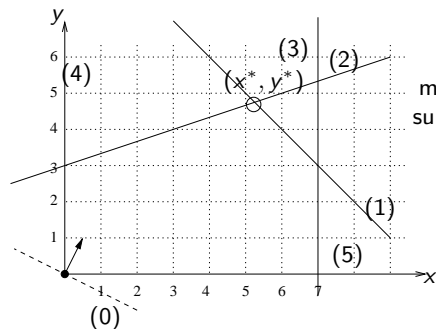
\Rightarrow New set of constraints:

$$\begin{bmatrix} x_1 + 2x_2 & -My \leq 4 \\ 5x_1 + 3x_2 - M(1 - y) \leq 10 \\ & y \in \{0, 1\} \\ & x_1, x_2 \geq 0 \end{bmatrix}$$

- $y = \begin{cases} 0 & \Rightarrow x_1 + 2x_2 \leq 4 \text{ must hold} \\ 1 & \Rightarrow 5x_1 + 3x_2 \leq 10 \text{ must hold} \end{cases}$

- 1 Suppose that you are interested in choosing from a set of investments $\{1, \dots, 7\}$ using 0/1 variables. Model the following constraints:
 - 1 You cannot invest in all of them
 - 2 You must choose at least one of them
 - 3 Investment 1 cannot be chosen if investment 3 is chosen
 - 4 Investment 4 can be chosen only if investment 2 is also chosen
 - 5 You must choose either both investment 1 and 5 or neither
 - 6 You must choose either at least one of the investments 1, 2 and 3 or at least two investments from 2, 4, 5 and 6
- 2 Formulate the following as mixed integer programs:
 - 1 $u = \min\{x_1, x_2\}$, assuming that $0 \leq x_j \leq C$ for $j = 1, 2$
 - 2 $v = |x_1 - x_2|$ with $0 \leq x_j \leq C$ for $j = 1, 2$
 - 3 The set $X \setminus \{x^*\}$ where $X = \{x \in \mathbb{Z}^n \mid Ax \leq b\}$ and $x^* \in X$

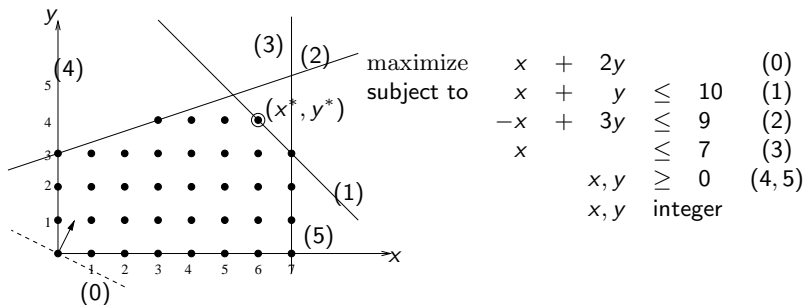
Linear programming: A small example



$$\begin{array}{rcll} \text{maximize} & x & + & 2y & (0) \\ \text{subject to} & x & + & y & \leq 10 & (1) \\ & -x & + & 3y & \leq 9 & (2) \\ & x & & & \leq 7 & (3) \\ & & & x, y & \geq 0 & (4,5) \end{array}$$

- Optimal solution: $(x^*, y^*) = (5\frac{1}{4}, 4\frac{3}{4})$
- Optimal objective value: $14\frac{3}{4}$

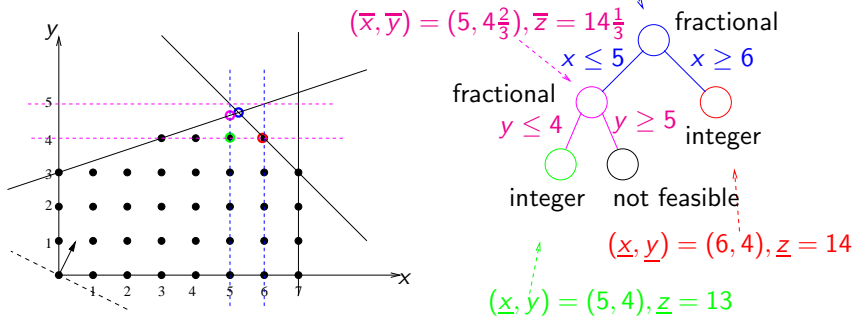
Integer linear programming: A small example



- What if the variables must take integer values?
- Optimal solution: $(x^*, y^*) = (6, 4)$
- Optimal objective value: $14 < 14\frac{3}{4}$
- The optimal value decreases (possibly constant) when the variables are restricted to possess only integral values

ILP: Solution by the branch-and-bound algorithm (e.g., Gurobi, Cplex, or CLP) (Ch. 15.1–15.2)

- Relax integrality requirements \Rightarrow linear, continuous problem $\Rightarrow (\bar{x}, \bar{y}) = (5\frac{1}{4}, 4\frac{3}{4}), \bar{z} = 14\frac{3}{4}$
- Search tree: branch on fractional variable values



For n binary variables: $\leq 2^n$ branches in the search tree

- Select an optimal collection of objects or investments or projects or ...
 - c_j = benefit of choosing object j , $j = 1, \dots, n$
- Limits on the budget
 - a_j = cost of object j , $j = 1, \dots, n$
 - b = total budget

- Variables: $x_j = \begin{cases} 1, & \text{if object } j \text{ is chosen,} \\ 0, & \text{otherwise,} \end{cases} \quad j = 1, \dots, n$

- Objective function:

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

- Budget constraint:

$$\sum_{j=1}^n a_j x_j \leq b$$

- Binary variables:

$$x_j \in \{0, 1\}, \quad j = 1, \dots, n$$

Computational complexity—the knapsack problem (Ch 2.6)

A small knapsack instance

$$\begin{aligned} z_1^* = \max \quad & 213x_1 + 1928x_2 + 11111x_3 + 2345x_4 + 9123x_5 \\ \text{subject to} \quad & 12223x_1 + 12224x_2 + 36674x_3 + 61119x_4 + 85569x_5 \leq 89\,643\,482 \\ & x_1, \dots, x_5 \geq 0, \text{ integer} \end{aligned}$$

- Optimal solution $\mathbf{x}^* = (0, 1, 2444, 0, 0)$, $z_1^* = 27\,157\,212$
- Cplex finds this solution in 0.015 seconds

The equality version

$$\begin{aligned} z_2^* = \max \quad & 213x_1 + 1928x_2 + 11111x_3 + 2345x_4 + 9123x_5 \\ \text{subject to} \quad & 12223x_1 + 12224x_2 + 36674x_3 + 61119x_4 + 85569x_5 = 89\,643\,482 \\ & x_1, \dots, x_5 \geq 0, \text{ integer} \end{aligned}$$

- Optimal solution $\mathbf{x}^* = (7334, 0, 0, 0, 0)$, $z_2^* = 1\,562\,142$
- Cplex computations interrupted after 1700 sec. ($\approx \frac{1}{2}$ hour)
 - No integer solution found
 - Best upper bound found: 25 821 000
 - 55 863 802 branch-and-bound nodes visited
 - Only *one* feasible solution exists!

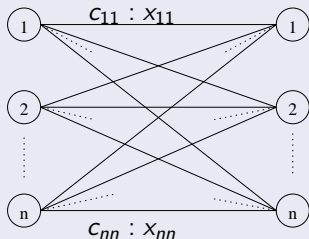
Assign each task to one resource, and each resource to one task

- A cost c_{ij} for assigning task i to resource j , $i, j \in \{1, \dots, n\}$
- Variables: $x_{ij} = \begin{cases} 1, & \text{if task } i \text{ is assigned to resource } j \\ 0, & \text{otherwise} \end{cases}$

$$\begin{aligned} \min \quad & \sum_{i=1}^n \sum_{j=1}^n c_{ij} x_{ij} \\ \text{subject to} \quad & \sum_{j=1}^n x_{ij} = 1, \quad i = 1, \dots, n \\ & \sum_{i=1}^n x_{ij} = 1, \quad j = 1, \dots, n \\ & x_{ij} \geq 0, \quad i, j = 1, \dots, n \end{aligned}$$

The assignment model

Choose *one* element from each row and each column



c_{11}	c_{12}	c_{13}					c_{1n}
c_{21}	c_{22}	c_{23}					c_{2n}
c_{31}	c_{32}	c_{33}					c_{3n}
c_{n1}	c_{n2}	c_{n3}					c_{nn}

- This integer linear model has *integral extreme points*, since it can be formulated as a network flow problem (Lect. 11–12)
- Therefore, it can be efficiently solved using specialized (network) linear programming techniques
- Even more efficient special purpose (primal–dual–graph-based) algorithms exist

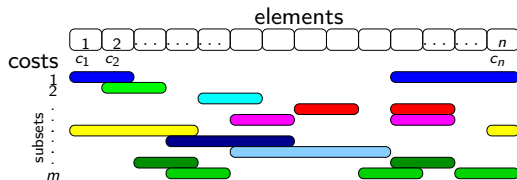
Computational complexity

- Mathematical insight yields efficient algorithms
- E.g., the *assignment problem*
 - # feasible solutions: $n! \implies$ Combinatorial explosion
 - An algorithm \exists that solves this problem in time $\mathcal{O}(n^4) \propto n^4$

Complete enumeration of all solutions is *not* efficient

n	2	5	8	10	100	1000
$n!$	2	120	40 000	3 600 000	$9.3 \cdot 10^{157}$	$4.0 \cdot 10^{2567}$
2^n	4	32	256	1 024	$1.3 \cdot 10^{30}$	$1.1 \cdot 10^{301}$
n^4	16	625	4 100	10 000	$1.0 \cdot 10^8$	$1.0 \cdot 10^{12}$
$n \log n$	0.6	3.5	7.2	10	200	3 000

- Binary knapsack: $\mathcal{O}(2^n)$
- Continuous knapsack (sorting of $\frac{c_j}{a_j}$): $\mathcal{O}(n \log n)$



Mathematical formulation

$$\begin{array}{ll}
 \min & \mathbf{c}^T \mathbf{x} \\
 \text{subject to} & \mathbf{A}\mathbf{x} \geq \mathbf{1} \\
 & \mathbf{x} \text{ binary}
 \end{array}$$

- $\mathbf{c} \in \mathbb{R}^n$ and $\mathbf{1} = (1, \dots, 1)^T \in \mathbb{R}^m$ are constant vectors
- $\mathbf{A} \in \mathbb{R}^{m \times n}$ is a matrix with entries $a_{ij} \in \{0, 1\}$
- $\mathbf{x} \in \mathbb{R}^n$ is the vector of variables
- Related models: *set partitioning* ($\mathbf{A}\mathbf{x} = \mathbf{1}$), *set packing* ($\mathbf{A}\mathbf{x} \leq \mathbf{1}$)