

MVE165/MMG631

Linear and integer optimization with applications

Lecture 7

Discrete optimization models and applications;  
complexity

Ann-Brith Strömberg

2018-04-17

# 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

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

- What if we are allowed to buy at most  $k$  different kinds of food, where  $k \leq n$ ?
- 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

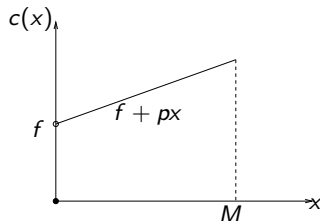
- *Linear programming* (LP) uses continuous variables:  $x_{ij} \geq 0$
- *Integer linear programming* (ILP) uses *integer* variables:  $x_{ij} \in \mathbb{Z}$
- *Binary linear programming* (BLP) uses *binary* variables:  $x_{ij} \in \mathbb{B}$
- If *both* continuous and integer/binary variables are 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:

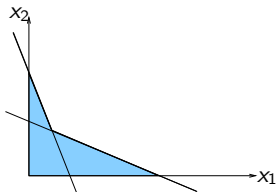
$$\left[ \begin{array}{ll} \min & fy + px \\ \text{s.t.} & x - My \leq 0 \\ & x \geq 0 \\ & y \in \{0, 1\} \end{array} \right]$$

- $y = 0 \Rightarrow x = 0 \Rightarrow fy + px = 0$
- $y = 1 \Rightarrow x \leq M \Rightarrow fy + px = f + px$
- $x > 0 \Rightarrow y = 1 \Rightarrow fy + px = f + px$
- $x = 0 \not\Rightarrow y = 0$  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  $y \in \{0, 1\}$

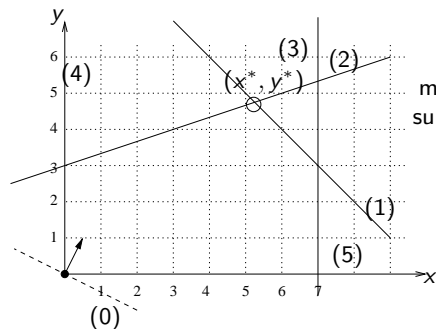
$\Rightarrow$  New set of constraints:

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

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

- 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  $x^* \in X$

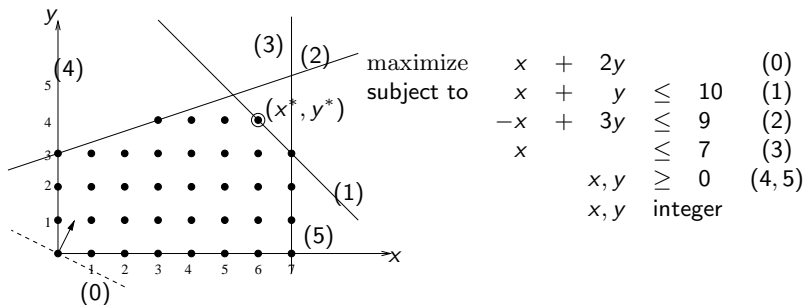
# Linear programming: A small example



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

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

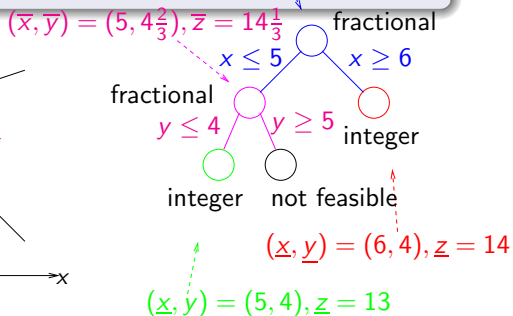
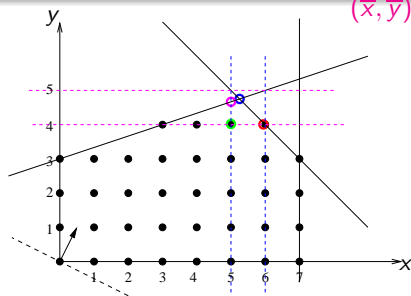
# Integer linear programming: A small example



- 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., Cplex, XpressMP, or GLPK) (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 over 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

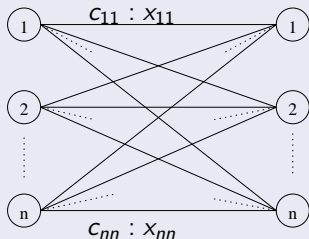
- A cost  $c_{ij}$  for assigning task  $i$  to resource  $j$ ,  $i, j \in \{1, \dots, n\}$
- Variables:  $x_{ij} = \begin{cases} 1, & \text{if task } i \text{ is assigned to resource } j \\ 0, & \text{otherwise} \end{cases}$

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



# 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  $\mathcal{O}(n^4) \propto n^4$

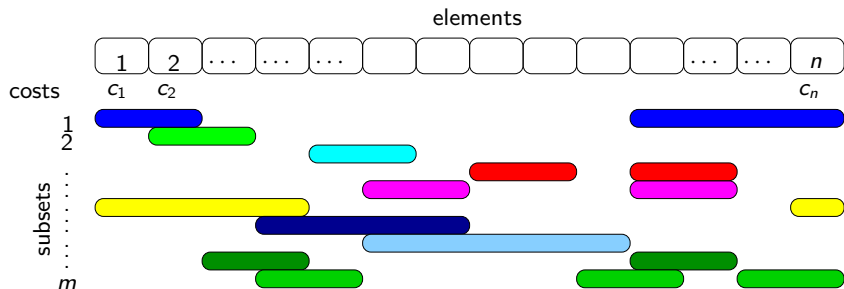
Complete enumeration of all solutions is *not* efficient

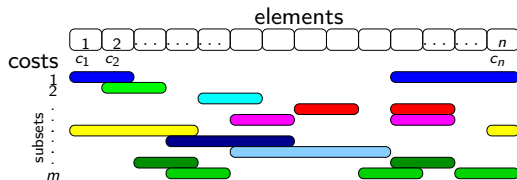
$n$	2	5	8	10	100	1000
$n!$	2	120	40 000	3 600 000	$9.3 \cdot 10^{157}$	$4.0 \cdot 10^{2567}$
$2^n$	4	32	256	1 024	$1.3 \cdot 10^{30}$	$1.1 \cdot 10^{301}$
$n^4$	16	625	4 100	10 000	$1.0 \cdot 10^8$	$1.0 \cdot 10^{12}$
$n \log n$	0.6	3.5	7.2	10	200	3 000

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

# Set covering problem—exponential complexity (Ch. 13.8)

- A number ( $n$ ) of items and a cost for each item
- A number ( $m$ ) of subsets of the  $n$  items
- Find a selection of the items such that each subset contains at least one selected item and such that the total cost for the selected items is minimized





## 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* ( $\mathbf{A} \mathbf{x} \leq \mathbf{1}$ )