

DM204, 2010
SCHEDULING, TIMETABLING AND ROUTING

Lecture 3
Resource Constrained Project Scheduling
Linear and Integer Programming (1)

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1. Scheduling

Resource Constrained Project Scheduling Model

2. Mathematical Programming

Introduction

Solution Algorithms

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Introduction

Solution Algorithms

Given:

- activities (jobs) $j = 1, \dots, n$
- renewable resources $i = 1, \dots, m$
- amount of resources available R_i
- processing times p_j
- amount of resource used r_{ij}
- precedence constraints $j \rightarrow k$

Further generalizations

- Time dependent resource profile $R_i(t)$
given by (t_i^μ, R_i^μ) where $0 = t_i^1 < t_i^2 < \dots < t_i^{m_i} = T$
Disjunctive resource, if $R_k(t) = \{0, 1\}$; cumulative resource, otherwise
- Multiple modes for an activity j
processing time and use of resource depends on its mode m : p_{jm}, r_{jkm} .

Assignment 1

- A contractor has to complete n activities.
- The duration of activity j is p_j
- each activity requires a crew of size W_j .
- The activities are not subject to precedence constraints.
- The contractor has W workers at his disposal
- his objective is to complete all n activities in minimum time.

Assignment 2

- Exams in a college may have different duration.
- The exams have to be held in a gym with W seats.
- The enrollment in course j is W_j and
- all W_j students have to take the exam at the same time.
- The goal is to develop a timetable that schedules all n exams in minimum time.
- Consider both the cases in which each student has to attend a single exam as well as the situation in which a student can attend more than one exam.

Assignment 3

- In a basic high-school timetabling problem we are given m classes c_1, \dots, c_m ,
- h teachers a_1, \dots, a_h and
- T teaching periods t_1, \dots, t_T .
- Furthermore, we have lectures $i = l_1, \dots, l_n$.
- Associated with each lecture is a unique teacher and a unique class.
- A teacher a_j may be available only in certain teaching periods.
- The corresponding timetabling problem is to assign the lectures to the teaching periods such that
 - each class has at most one lecture in any time period
 - each teacher has at most one lecture in any time period,
 - each teacher has only to teach in time periods where he is available.

Assignment 4

- A set of jobs J_1, \dots, J_g are to be processed by auditors A_1, \dots, A_m .
- Job J_l consists of n_l tasks ($l = 1, \dots, g$).
- There are precedence constraints $i_1 \rightarrow i_2$ between tasks i_1, i_2 of the same job.
- Each job J_l has a release time r_l , a due date d_l and a weight w_l .
- Each task must be processed by exactly one auditor. If task i is processed by auditor A_k , then its processing time is p_{ik} .
- Auditor A_k is available during disjoint time intervals $[s_k^\nu, l_k^\nu]$ ($\nu = 1, \dots, m$) with $l_k^\nu < s_k^{\nu+1}$ for $\nu = 1, \dots, m_k - 1$.
- Furthermore, the total working time of A_k is bounded from below by H_k^- and from above by H_k^+ with $H_k^- \leq H_k^+$ ($k = 1, \dots, m$).
- We have to find an assignment $\alpha(i)$ for each task $i = 1, \dots, n := \sum_{l=1}^g n_l$ to an auditor $A_{\alpha(i)}$ such that
 - each task is processed without preemption in a time window of the assigned auditor
 - the total workload of A_k is bounded by H_k^- and H_k^+ for $k = 1, \dots, m$.
 - the precedence constraints are satisfied,
 - all tasks of J_l do not start before time r_l , and
 - the total weighted tardiness $\sum_{l=1}^g w_l T_l$ is minimized.

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program = optimization problem

$$\begin{aligned} \min \quad & f(x) \\ & g_i(x) = 0, \quad i = 1, 2, \dots, k \\ & h_j(x) \leq 0, \quad j = 1, 2, \dots, m \\ & x \in \mathbf{R}^n \end{aligned}$$

general (nonlinear) program (NLP)

$$\begin{aligned} \min \quad & c^T x \\ & Ax = a \\ & Bx \leq b \\ & x \geq 0 \\ & (x \in \mathbf{R}^n) \end{aligned}$$

linear program (LP)

$$\begin{aligned} \min \quad & c^T x \\ & Ax = a \\ & Bx \leq b \\ & x \geq 0 \\ & (x \in \mathbf{Z}^n) \\ & (x \in \{0, 1\}^n) \end{aligned}$$

integer (linear) program (IP, MIP)

Linear Program in standard form

$$\begin{aligned} \min \quad & c_1x_1 + c_2x_2 + \dots + c_nx_n \\ \text{s.t.} \quad & a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n = b_1 \\ & a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n = b_2 \\ & \vdots \\ & a_{n1}x_1 + a_{n2}x_2 + \dots + a_{nn}x_n = b_n \\ & x_1, x_2, \dots, x_n \geq 0 \end{aligned}$$

$$\begin{aligned} \min \quad & c^T x \\ & Ax = b \\ & x \geq 0 \end{aligned}$$

- 1939 L. V. Kantorovitch: Foundations of linear programming (Nobel Prize 1975)
- George J. Stigler's 1945 (Nobel Prize 1982) "Diet Problem": "the first linear program"
find the cheapest combination of foods that will satisfy the daily requirements of a person
Army's problem had 77 unknowns and 9 constraints.
<http://www.mcs.anl.gov/home/otc/Guide/CaseStudies/diet/index.html>
- 1947 G.B. Dantzig: Invention of the simplex algorithm
- Founding fathers:
 - 1950s Dantzig: Linear Programming 1954, the Beginning of IP
G. Dantzig, D.R. Fulkerson, S. Johnson TSP with 49 cities
 - 1960s Gomory: Integer Programming

- Max-Flow Min-Cut Theorem

The maximal (s,t)-flow in a capacitated network is equal to the minimal capacity of an (s,t)-cut

- The Duality Theorem of Linear Programming

$$\begin{aligned} \max \quad & c^T x \\ & Ax \leq b \\ & x \geq 0 \end{aligned}$$

$$\begin{aligned} \min \quad & y^T b \\ & y^T A \geq c^T \\ & y \geq 0 \end{aligned}$$

If feasible solutions to both the primal and the dual problem in a pair of dual LP problems exist, then there is an optimum solution to both systems and the optimal values are equal.

- **Max-Flow Min-Cut Theorem**
does not hold if several source-sink relations are given
(multicommodity flow)
- **The Duality Theorem of Integer Programming**

$$\begin{array}{ll} \max & c^T x \\ & Ax \leq b \\ & x \geq 0 \\ & x \in \mathbf{Z}^n \end{array} \leq \begin{array}{ll} \min & y^T b \\ & y^T A \geq c^T \\ & y \geq 0 \\ & y \in \mathbf{Z}^n \end{array}$$

- Linear programs can be solved in polynomial time with the Ellipsoid Method (Khachiyan, 1979)
Interior Point Methods (Karmarkar, 1984, and others)
- **Open:** is there a strongly polynomial time algorithm for the solution of LPs?
- Certain variants of the Simplex Algorithm run - under certain conditions - in expected polynomial time (Borgwardt, 1977...)
- **Open:** Is there a polynomial time variant of the Simplex Algorithm?

Theorem (Grötschel, Lovász, Schrijver 1979, 1988))

(modulo technical details)

There exists a polynomial time algorithm to minimize convex functions (e.g., linear functions) over the elements of a class of convex bodies \mathbf{K} (e. g. polyhedra) if and only if, there exists a polynomial time algorithm that decides, for any given point x , whether x is in \mathbf{K} , and that, when x is not in \mathbf{K} , finds a hyperplane that separates x from \mathbf{K} .

Short version: Optimization and Separation are polynomial-time equivalent.

Consequence: Theoretical Foundation of cutting plane algorithms.

- **Theorem**
Integer, 0/1, and mixed integer programming are NP-hard.
- **Consequence**
 - investigate special cases
 - special purpose algorithms
 - heuristics

- Algorithms for the solution of nonlinear programs
- Algorithms for the solution of linear programs
 - 1) Fourier-Motzkin Elimination (hopeless)
 - 2) The Simplex Method (good, above all with duality)
 - 3) The Ellipsoid Method (total failure)
 - 4) Interior-Point/Barrier Methods (good)
- Algorithms for the solution of integer programs
 - 1) Branch & Bound
 - 2) Cutting Planes

- Iterative methods that solve the equation and inequality systems representing the necessary local optimality conditions (Kuhn-Tucker sufficient conditions)
- Steepest descent
- Newton method
- Subgradient method (for piecewise functions)

The Simplex Method

- Dantzig, 1947: primal Simplex Method
- Lemke, 1954; Beale, 1954: dual Simplex Method
- Dantzig, 1953: revised Simplex Method
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- Underlying Idea: Find a vertex of the set of feasible LP solutions (polyhedron) and move to a better neighbouring vertex, if possible.

The simplex method

min/max $x_1 + 3x_2$

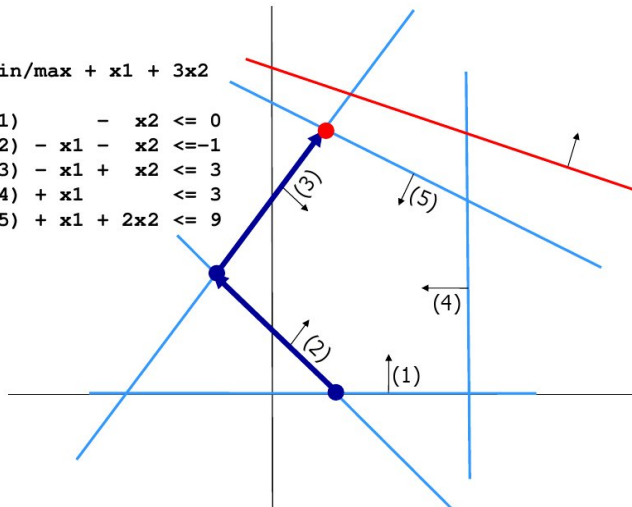
(1) $-x_2 \leq 0$

(2) $-x_1 - x_2 \leq -1$

(3) $-x_1 + x_2 \leq 3$

(4) $+x_1 \leq 3$

(5) $+x_1 + 2x_2 \leq 9$



The simplex method

min/max $+ x_1 + 3x_2$

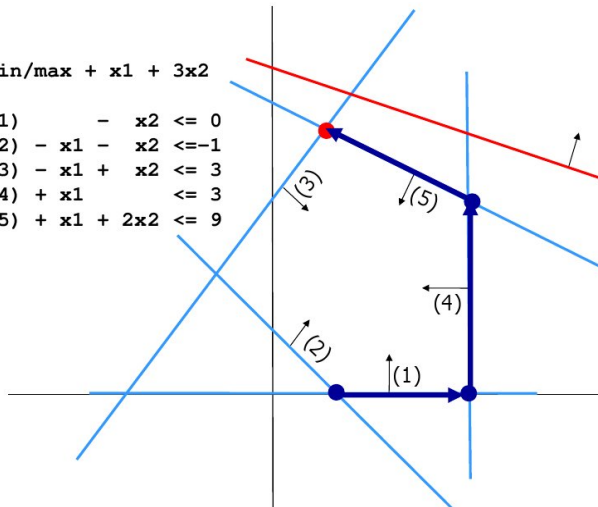
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(5) $+ x_1 + 2x_2 \leq 9$



Hirsch Conjecture

If P is a polytope of dimension n with m facets then every vertex of P can be reached from any other vertex of P on a path of length at most $m - n$.

In the example before: $m = 5$, $n = 2$ and $m - n = 3$, conjecture is true.

At present, not even a polynomial bound on the path length is known.

Best upper bound: Kalai, Kleitman (1992): The diameter of the graph of an n -dimensional polyhedron with m facets is at most $m^{(\log n + 1)}$.

Lower bound: Holt, Klee (1997): at least $m - n$ (m, n large enough).

special „simple“ combinatorial optimization problems Finding a:

- minimum spanning tree
- shortest path
- maximum matching
- maximal flow through a network
- cost-minimal flow
- ...

solvable in polynomial time by special purpose algorithms

special „hard“ combinatorial optimization problems

- traveling salesman problem
- location and routing
- set-packing, partitioning, -covering
- max-cut
- linear ordering
- scheduling (with a few exceptions)
- node and edge colouring
- ...

NP-hard (in the sense of complexity theory)

The most successful solution techniques employ linear programming.

- 1) Branch & Bound
- 2) Cutting Planes

Branch & cut, Branch & Price (column generation), Branch & Cut & Price

- We can solve today **explicit LPs** with
 - up to 500,000 of variables and
 - up to 5,000,000 of constraintsin relatively short running times.
- We can solve today structured **implicit LPs** (employing column generation and cutting plane techniques) in special cases with hundreds of million (and more) variables and almost infinitely many constraints in acceptable running times. (Examples: TSP, bus circulation in Berlin)

[Martin Grötschel, Block Course at TU Berlin,
“Combinatorial Optimization at Work”, 2005
<http://co-at-work.zib.de/berlin/>]