DM545/DM871
Linear and Integer Programming

Lecture 9
Integer Linear Programming
Modeling

Marco Chiarandini

Department of Mathematics & Computer Science
University of Southern Denmark

Midway evaluation

What is working well and what is not:

https://padlet.com/marco43/dm545_2020

https://padlet.com/marco43/dm871_2020

Outline

Outline

Discrete Optimization

- Often we need to deal with integral inseparable quantities

- Sometimes rounding can go

- Other times rounding not feasible: eg, presence of a bus on a line is 0.3...

Integer Linear Programming

Linear Objective
Linear Constraints
but! integer variables

The world is not linear: "OR is the art and science of obtaining bad answers to questions to which otherwise worse answers would be given"

$$\max \mathbf{c}^T\mathbf{x}$$
$$A\mathbf{x} \leq \mathbf{b}$$
$$\mathbf{x} \geq \mathbf{0}$$

Linear Programming (LP)

$$\max \mathbf{c}^T\mathbf{x}$$
$$A\mathbf{x} \leq \mathbf{b}$$
$$\mathbf{x} \geq \mathbf{0}$$
$$\mathbf{x} \text{ integer}$$

Integer (Linear) Programming (ILP)

$$\max \mathbf{c}^T\mathbf{x}$$
$$A\mathbf{x} \leq \mathbf{b}$$
$$\mathbf{x} \in \{0,1\}^n$$

Binary Integer Program (BIP)
0/1 Integer Programming

$$\max \mathbf{c}^T\mathbf{x} + \mathbf{h}^T\mathbf{y}$$
$$A\mathbf{x} + G\mathbf{y} \leq \mathbf{b}$$
$$\mathbf{x} \geq \mathbf{0}$$
$$\mathbf{y} \geq \mathbf{0}$$
$$\mathbf{y} \text{ integer}$$

Mixed Integer (Linear) Programming (MILP)

$$\max f(\mathbf{x})$$
$$g(\mathbf{x}) \leq \mathbf{b}$$
$$\mathbf{x} \geq \mathbf{0}$$
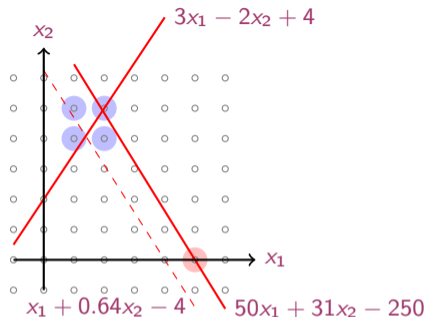
Non-linear Programming (NLP)

Recall:

- $\mathbb{Z}$ set of integers

- $\mathbb{Z}^+$ set of positive integer

- $\mathbb{Z}_0^+$ set of nonnegative integers ($\{0\} \cup \mathbb{Z}^+$)

- $\mathbb{N}_0$ set of natural numbers, ie, nonnegative integers $\{0, 1, 2, 3, 4, ...\}$

Rounding

$$\max 100x_1 + 64x_2$$
$$50x_1 + 31x_2 \leq 250$$
$$3x_1 - 2x_2 \geq -4$$
$$x_1, x_2 \in \mathbb{Z}_0^+$$

Note: rounding does not help in the example above!

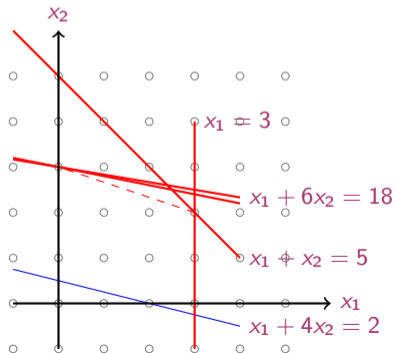LP optimum $(376/193, 950/193)$
IP optimum $(5, 0)$

$\rightsquigarrow$ feasible region convex but not continuous: Now the optimum can be on the border (vertices) but also internal.

Possible way: solve the relaxed problem.
- If solution is integer, done.
- If solution is rational (never irrational) try rounding to the nearest integers (but may exit feasibility region)
  if in $\mathbb{R}^2$ then 2^2 possible roundings (up or down)
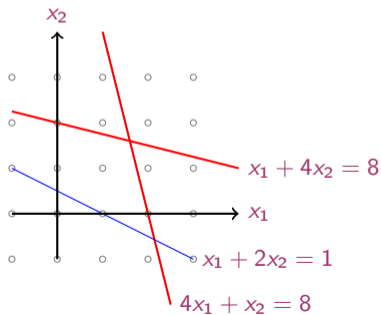 if in $\mathbb{R}^n$ then 2^n possible roundings (up or down)

Cutting Planes

$$\max x_1 + 4x_2$$
$$x_1 + 6x_2 \leq 18$$
$$x_1 \qquad \leq 3$$
$$x_1, x_2 \geq 0$$
$$x_1, x_2 \text{ integers}$$

Branch and Bound

$$\max \ x_1 \ + 2x_2$$
$$x_1 \ + 4x_2 \leq 8$$
$$4x_1 + \ x_2 \ \leq 8$$
$$x_1, x_2 \geq 0, \text{integer}$$
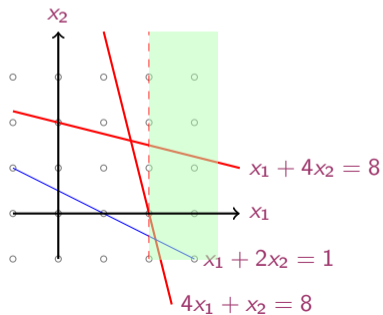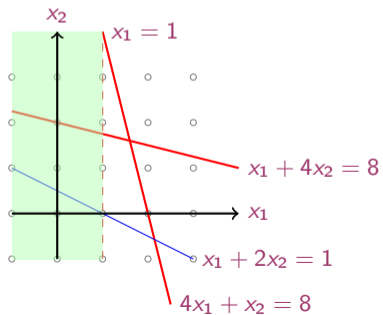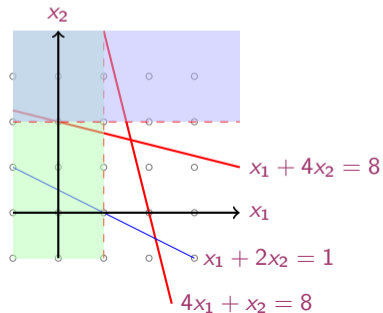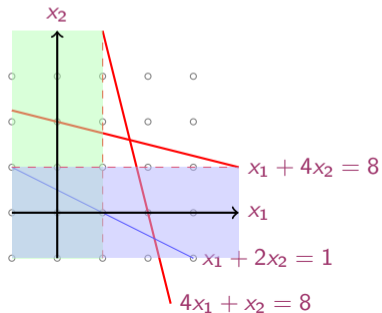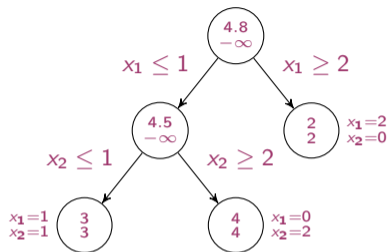
$x_1 + 4x_2 = 8$

$x_1 + 2x_2 = 1$

$4x_1 + x_2 = 8$

Outline

Mathematical Programming: Modeling

- Find out exactly what the decision maker 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 and Known Parameters corresponding to given data.
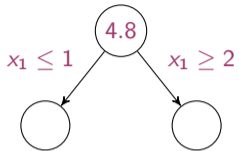
- Formulate Objective Function computing the benefit/cost

- Formulate mathematical Constraints indicating the interplay between the different variables.

How to "build" a constraint

- Formulate relationship between the variables in plain words
- Then formulate your sentences using logical connectives **and, or, not, implies**
- Finally convert the logical statement to a mathematical constraint.

Example

- "The power plant must not work in both of two neighbouring time periods"
- on/off is modelled using **binary** integer variables
- $x_i = 1$ or $x_i = 0$
- $x_i = 1$ implies $\Rightarrow x_{i+1} = 0$
- $x_i + x_{i+1} \leq 1$

Outline

The Assignment Problem

Problem

Common application: **Assignees** are being assigned to perform **tasks**.

Suppose we have n persons and n jobs
Each person has a certain proficiency at each job.

Formulate a mathematical model that can be used to find an assignment that maximizes the total proficiency.

The Assignment Problem
Model

Decision Variables:

$$x_{ij} = \begin{cases} 1 \text{ if person } i \text{ is assigned job } j \\ 0 \text{ otherwise,} \end{cases} \text{ for } i, j = 1, 2, \ldots, n$$

Objective Function:

$$\max \quad \sum_{i=1}^{n} \sum_{j=1}^{n} \rho_{ij} x_{ij}$$

where ρ_{ij} is person i's proficiency at job j

The Assignment Problem
Model

Constraints:
Each person is assigned one job:

$$\sum_{j=1}^{n} x_{ij} = 1 \text{ for all } i$$

e.g. for person 1 we get $x_{11} + x_{12} + x_{13} + \cdots + x_{1n} = 1$

Each job is assigned to one person:

$$\sum_{i=1}^{n} x_{ij} = 1 \text{ for all } j$$

e.g. for job 1 we get $x_{11} + x_{21} + x_{31} + \cdots + x_{n1} = 1$

Outline

The Knapsack Problem

Input: Given a set of n items, each with a value v_i and weight w_i ($i = 1, \ldots, n$)

Task: determine the number of each items to include in a collection so that the total weight is less than a given limit, W, and the total value is as large as possible.

The "knapsack" name derives from the problem faced by someone who is constrained by a fixed-size knapsack and must fill it with the most useful items.

Assuming we can take at most one of any item and that $\sum_i w_i > W$, formulate a mathematical model to determine which items give the largest value.

Model used, eg, in capital budgeting, project selection, etc.

The Knapsack Problem

Decision Variables:

$$x_i = \begin{cases} 1 \text{ if item } i \text{ is taken} \\ 0 \text{ otherwise,} \end{cases} \text{ for } i = 1, 2 \ldots, n$$

Objective Function:

$$\max \quad \sum_{i=1}^{n} v_i x_i$$

Constraints:

Knapsack capacity restriction:

$$\sum_{i=1}^{n} w_i x_i \leq W$$

Outline

Set Covering

Problem

Given: a set of regions, a set of possible construction locations for emergency centers, regions that can be served in less than 8 minutes, cost of installing an emergency center in each location.

Task: decide where to install a set of emergency centers such that the total cost is minimized and all regions are safely served

As a COP: $M = \{1, \ldots, m\}$ regions, $\quad N = \{1, \ldots, n\}$ centers, $\quad S_j \subseteq M$ regions serviced by $j \in N$ in 8 min.

$$\min_{T \subseteq N} \left\{ \sum_{j \in T} c_j \mid \bigcup_{j \in T} S_j = M \right\}$$

regions: $M = \{1, \ldots, 5\}$
centers: $N = \{1, \ldots, 6\}$
cost of centers: $c_j = 1 \quad \forall j = 1, \ldots, 6$
coverages: $S_1 = (1, 2), S_2 = (1, 3, 5), S_3 = (2, 4, 5), S_4 = (3), S_5 = (1), S_6 = (4, 5)$

Example

- regions: $M = \{1, \ldots, 5\}$
  centers: $N = \{1, \ldots, 6\}$
  cost of centers: $c_j = 1 \quad \forall j = 1, \ldots, 6$
  coverages: $S_1 = (1, 2), S_2 = (1, 3, 5), S_3 = (2, 4, 5), S_4 = (3), S_5 = (1), S_6 = (4, 5)$

-

$$
A = \begin{array}{c} \\ \\ 1 \\ 2 \\ 3 \\ 4 \\ 5 \end{array}
\begin{array}{c} x_1 \\ S_1 \\ \left[\begin{array}{c} 1 \\ 1 \\ 0 \\ 0 \\ 0 \end{array} \right. \end{array}
\begin{array}{c} x_2 \\ S_2 \\ 1 \\ 0 \\ 1 \\ 0 \\ 1 \end{array}
\begin{array}{c} x_3 \\ S_3 \\ 0 \\ 1 \\ 0 \\ 1 \\ 1 \end{array}
\begin{array}{c} x_4 \\ S_4 \\ 0 \\ 0 \\ 1 \\ 0 \\ 0 \end{array}
\begin{array}{c} x_5 \\ S_5 \\ 1 \\ 0 \\ 0 \\ 0 \\ 0 \end{array}
\begin{array}{c} x_6 \\ S_6 \\ \left. \begin{array}{c} 0 \\ 0 \\ 0 \\ 1 \\ 1 \end{array} \right] \end{array}
$$

As a BIP:

Variables:
$\mathbf{x} \in \mathbb{B}^n$, $x_j = 1$ if center j is selected, 0 otherwise

Objective:

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

Constraints:

- incidence matrix: $a_{ij} = \begin{cases} 1 \\ 0 \end{cases}$

- $\sum_{j=1}^n a_{ij} x_j \geq 1$

Set covering
cover each of M at least once

1. min, $\geq$
2. all RHS terms are 1
3. all matrix elements are 1

$$\min \mathbf{c}^T \mathbf{x}$$
$$A\mathbf{x} \geq \mathbf{1}$$
$$\mathbf{x} \in \mathbb{B}^n$$

Set packing
cover as many of M without overlap

1. max, $\leq$
2. all RHS terms are 1
3. all matrix elements are 1

$$\max \mathbf{c}^T \mathbf{x}$$
$$A\mathbf{x} \leq \mathbf{1}$$
$$\mathbf{x} \in \mathbb{B}^n$$

Set partitioning
cover exactly once each element of M

1. max or min, $=$
2. all RHS terms are 1
3. all matrix elements are 1

$$\max \mathbf{c}^T \mathbf{x}$$
$$A\mathbf{x} = \mathbf{1}$$
$$\mathbf{x} \in \mathbb{B}^n$$

Generalization: $RHS \geq 1$
Application examples:

- Aircrew scheduling: M: legs to cover, N: rosters
- Vehicle routing: M: customers, N: routes

A good written example of how to present a model:

2.1. Notation

Let N be the set of operational flight legs and K the set of aircraft types. Denote by n^k the number of available aircraft of type $k \in K$. Define Ω^k, indexed by p, as the set of feasible schedules for aircraft of type $k \in K$ and let index $p = 0$ denote the empty schedule for an aircraft. Next associate with each schedule $p \in \Omega^k$ the value c_p^k denoting the anticipated profit if this schedule is assigned to an aircraft of type $k \in K$ and a_{ip}^k a binary constant equal to 1 if this schedule covers flight leg $i \in N$ and 0 otherwise. Furthermore, let S be the set of stations and $S^k \subseteq S$ the subset having the facilities to serve aircraft of type $k \in K$. Then, define o_{sp}^k and d_{sp}^k to equal to 1 if schedule p, $p \in \Omega^k$, starts and ends respectively at station s, $s \in S^k$, and 0 otherwise.

Denote by θ_p^k, $p \in \Omega^k \backslash \{0\}$, $k \in K$, the binary decision variable which takes the value 1 if schedule p is assigned to an aircraft of type k, and 0 otherwise. Finally, let θ_0^k, $k \in K$, be a nonnegative integer variable which gives the number of unused aircraft of type k.

2.2. Formulation

Using these definitions, the DARSP can be formulated as:

$$\text{Maximize} \sum_{k \in K} \sum_{p \in \Omega^k} c_p^k \theta_p^k \qquad (1)$$

subject to:

$$\sum_{k \in K} \sum_{p \in \Omega^k} a_{ip}^k \theta_p^k = 1 \quad \forall i \in N, \qquad (2)$$

$$\sum_{p \in \Omega^k} (d_{sp}^k - o_{sp}^k) \theta_p^k = 0 \quad \forall k \in K, \forall s \in S^k, \qquad (3)$$

$$\sum_{p \in \Omega^k} \theta_p^k = n^k \quad \forall k \in K, \qquad (4)$$

$$\theta_p^k \geq 0 \quad \forall k \in K, \forall p \in \Omega^k, \qquad (5)$$

$$\theta_p^k \text{ integer} \quad \forall k \in K, \forall p \in \Omega^k. \qquad (6)$$

The objective function (1) states that we wish to maximize the total anticipated profit. Constraints (2) require that each operational flight leg be covered exactly once. Constraints (3) correspond to the flow conservation constraints at the beginning and the end of the day at each station and for each aircraft type. Constraints (4) limit the number of aircraft of type $k \in K$ that can be used to the number available. Finally, constraints (5) and (6) state that the decision variables are nonnegative integers. This model is a Set Partitioning problem with additional constraints.

Outline

Outline

Matching

Definition (Matching Theory Terminology)

Matching: set of pairwise non adjacent edges

Covered (vertex): a vertex is covered by a matching M if it is incident to an edge in M

Perfect (matching): if M covers each vertex in G

Maximal (matching): if M cannot be extended any further

Maximum (matching): if M covers as many vertices as possible

Matchable (graph): if the graph G has a perfect matching

$$\max \sum_{v \in V} w_e x_e$$
$$\sum_{e \in E: v \in e} x_e \leq 1 \qquad \forall v \in V$$
$$x_e \in \{0, 1\} \ \forall e \in E$$

Special case: bipartite matching $\equiv$ assignment problems

Vertex Cover

Select a subset $S \subseteq V$ such that each edge has at least one end vertex in S.

$$\min \sum_{v \in V} x_v$$
$$x_v + x_u \geq 1 \qquad \forall u, v \in V, uv \in E$$
$$x_v \in \{0, 1\} \ \forall v \in V$$

Approximation algorithm: set S derived from the LP solution in this way:

$$S_{LP} = \{v \in V : x_v^* \geq 1/2\}$$

(it is a cover since $x_v^* + x_u^* \geq 1$ implies $x_v^* \geq 1/2$ or $x_u^* \geq 1/2$)

Proposition

The LP rounding approximation algorithm gives a 2-approximation: $|S_{LP}| \leq 2|S_{OPT}|$
(at most as bad as twice the optimal solution)

Proof: Let $\bar{x}$ be opt to IP. Then $\sum x_v^* \leq \sum \bar{x}_v$.
$|S_{LP}| = \sum_{v \in S_{LP}} 1 \leq \sum_{v \in V} 2x_v^*$ since $x_v^* \geq 1/2$ for each $v \in S_{LP}$
$|S_{LP}| \leq 2 \sum_{v \in V} x_v^* \leq 2 \sum_{v \in V} \bar{x}_v = 2|S_{OPT}|$

Maximum Independent Set

Find the largest subset $S \subseteq V$ such that the induced graph has no edges

$$\max \sum_{v \in V} x_v$$
$$x_v + x_u \leq 1 \qquad \forall u, v \in V, uv \in E$$
$$x_v = \{0, 1\} \; \forall v \in V$$

Optimal sol of LP relaxation sets $x_v = 1/2$ for all variables and has value $|V|/2$.

What is the value of an optimal IP solution of a complete graph?

LP relaxation gives an $O(n)$-approximation (almost useless)

Traveling Salesman Problem

- Find the cheapest movement for a drilling, welding, drawing, soldering arm as, for example, in a printed circuit board manufacturing process or car manufacturing process

- n locations, c_{ij} cost of travel

Variables:

$$x_{ij} = \begin{cases} 1 \\ 0 \end{cases}$$

Objective:

$$\sum_{i=1}^{n} \sum_{j=1}^{n} c_{ij} x_{ij}$$

Constraints:

- $$\sum_{j:j\neq i} x_{ij} = 1 \qquad\qquad \forall i = 1, \ldots, n$$

  $$\sum_{i:i\neq j} x_{ij} = 1 \qquad\qquad \forall j = 1, \ldots, n$$

- cut set constraints

  $$\sum_{i\in S}\sum_{j\notin S} x_{ij} \geq 1 \qquad\qquad \forall S \subset N, S \neq \emptyset$$

- subtour elimination constraints

  $$\sum_{i\in S}\sum_{j\in S} x_{ij} \leq |S| - 1 \qquad\qquad \forall S \subset N, 2 \leq |S| \leq n - 1$$

Outline

Modeling Tricks

Objective function and/or constraints do not appear to be linear?

- Absolute values
- Minimize the largest function value
- Maximize the smallest function value
- Constraints include variable division
- Constraints are either/or
- A variable must take one of several candidate values

Modeling: Absolute Values

$$\min \sum_{i=1}^{n} |f_i(\mathbf{x})|$$

$$\min \sum_{i=1}^{n} z_i$$
$$\text{s.t.} \quad z_i \geq f_i(\mathbf{x}) \quad i = 1..n$$
$$z_i \geq -f_i(\mathbf{x}) \quad i = 1..n$$
$$z_i \in \mathbb{R} \quad i = 1..n$$
$$\mathbf{x} \in \mathbb{R}^q$$

n additional variables and $2n$ additional constraints.

$$\min \sum_{i=1}^{n} (z_i^+ + z_i^-)$$
$$\text{s.t.} \quad f_i(\mathbf{x}) = z_i^+ - z_i^- \quad i = 1..n$$
$$z_i^+, z_i^- \geq 0 \quad i = 1..n$$
$$\mathbf{x} \in \mathbb{R}^q$$

$2n$ additional variables and n additional constraints.

Modeling: Minimax

Minimize the largest of a number of function values:

$$\min \quad \max\{f_1(\mathbf{x}), \ldots, f_n(\mathbf{x})\}$$

- Introduce an auxiliary variable z:
$$\min \quad z$$
$$\text{s. t. } f_1(\mathbf{x}) \leq z$$
$$f_2(\mathbf{x}) \leq z$$

Modeling: Divisions

Constraints include variable division:

- Constraint of the form

$$\frac{a_1 x + a_2 y + a_3 z}{d_1 x + d_2 y + d_3 z} \leq b$$

  - Rearrange:

    $$a_1 x + a_2 y + a_3 z \leq b(d_1 x + d_2 y + d_3 z)$$

    which gives:

    $$(a_1 - bd_1)x + (a_2 - bd_2)y + (a_3 - bd_3)z \leq 0$$

Modeling: "Either/Or Constraints"

In conventional mathematical models, the solution must satisfy all constraints.
Suppose that your constraints are "either/or":

$$a_1 x_1 + a_2 x_2 \leq b_1 \qquad \text{or}$$
$$d_1 x_1 + d_2 x_2 \leq b_2$$

Introduce new variable $y \in \{0, 1\}$ and a large number M:

$$a_1 x_1 + a_2 x_2 \leq b_1 + My \qquad\qquad \text{if } y = 0 \text{ then this is active}$$
$$d_1 x_1 + d_2 x_2 \leq b_2 + M(1 - y) \qquad\qquad \text{if } y = 1 \text{ then this is active}$$

Modeling: "Either/Or Constraints"

Binary integer programming allows to model alternative choices:

- Eg: 2 feasible regions, ie, disjunctive constraints, not possible in LP.
  introduce y auxiliary binary variable and M a big number:

$$Ax \leq b + My \qquad \text{if } y = 0 \text{ then this is active}$$
$$A'x \leq b' + M(1 - y) \qquad \text{if } y = 1 \text{ then this is active}$$

Modeling: "Either/Or Constraints"

Generally:

$$a_{11}x_1 + a_{12}x_2 + a_{13}x_3 + \ldots + a_{1m}x_m \leq d_1$$
$$a_{21}x_1 + a_{22}x_2 + a_{23}x_3 + \ldots + a_{2m}x_m \leq d_2$$
$$\vdots$$
$$a_{m1}x_1 + a_{N2}x_2 + a_{N3}x_3 + \ldots + a_{Nm}x_m \leq d_N$$

Exactly K of the N constraints must be satisfied.
Introduce binary variables $y_1, y_2, \ldots, y_N$ and a large number M

$$a_{11}x_1 + a_{12}x_2 + a_{13}x_3 + \ldots + a_{1m}x_m \leq d_1 + My_1$$
$$a_{21}x_1 + a_{22}x_2 + a_{23}x_3 + \ldots + a_{2m}x_m \leq d_2 + My_2$$
$$\vdots$$
$$a_{m1}x_1 + a_{N2}x_2 + a_{N3}x_3 + \ldots + a_{Nm}x_m \leq d_N + My_N$$

$$y_1 + y_2 + \ldots y_N = N - K$$

K of the y-variables are 0, so K constraints must be satisfied

Modeling: "Either/Or Constraints"

At least $h \le k$ of $\sum\limits_{j=1}^{n} a_{ij} x_j \le b_i$, $i = 1, \ldots, k$ must be satisfied

introduce y_i, $i = 1, \ldots, k$ auxiliary binary variables

$$\sum_{j=1}^{n} a_{ij} x_j \le b_i + M y_i$$

$$\sum_i y_i \le k - h$$

Modeling: "Possible Constraints Values"

A constraint must take on one of N given values:

$$a_1 x_1 + a_2 x_2 + a_3 x_3 + \ldots + a_m x_m = d_1 \text{ or}$$
$$a_1 x_1 + a_2 x_2 + a_3 x_3 + \ldots + a_m x_m = d_2 \text{ or}$$
$$\vdots$$
$$a_1 x_1 + a_2 x_2 + a_3 x_3 + \ldots + a_m x_m = d_N$$

Introduce binary variables $y_1, y_2, \ldots, y_N$:

$$a_1 x_1 + a_2 x_2 + a_3 x_3 + \ldots + a_m x_m = d_1 y_1 + d_2 y_2 + \ldots d_N y_N$$

$$y_1 + y_2 + \ldots y_N = 1$$

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Outline

Uncapacited Facility Location (UFL)

Given:

- depots $N = \{1, \ldots, n\}$
- clients $M = \{1, \ldots, m\}$
- f_j fixed cost to use depot j
- transport cost for all orders c_{ij}

Task: Which depots to open and which depots serve which client

Variables: $y_j = \begin{cases} 1 & \text{if depot opened} \\ 0 & \text{otherwise} \end{cases}$, $\quad x_{ij}$ fraction of demand of i satisfied by j

Objective:

$$\min \sum_{i \in M} \sum_{j \in N} c_{ij} x_{ij} + \sum_{j \in N} f_j y_j$$

Constraints:

$$\sum_{j=1}^{n} x_{ij} = 1 \qquad\qquad\qquad \forall i = 1, \ldots, m$$

$$\sum_{i \in M} x_{ij} \leq m y_j \qquad\qquad\qquad \forall j \in N$$

Outline

Good and Ideal Formulations

Definition (Formulation)

A polyhedron $P \subseteq \mathbb{R}^{n+p}$ is a formulation for a set $X \subseteq \mathbb{Z}^n \times \mathbb{R}^p$ if and only if $X = P \cap (\mathbb{Z}^n \times \mathbb{R}^p)$

That is, if it does not leave out any of the solutions of the feasible region X.

There are infinite formulations

Definition (Convex Hull)

Given a set $X \subseteq \mathbb{Z}^n$ the convex hull of X is defined as:

$$\text{conv}(X) = \Big\{ \mathbf{x} : \mathbf{x} = \sum_{i=1}^{t} \lambda_i \mathbf{x}^i, \qquad \sum_{i=1}^{t} \lambda_i = 1, \qquad \lambda_i \geq 0, \qquad \text{for } i = 1, \ldots, t,$$
$$\text{for all finite subsets } \{\mathbf{x}^1, \ldots, \mathbf{x}^t\} \text{ of } X \Big\}$$

Proposition

conv(X) is a polyhedron (ie, representable as $A\mathbf{x} \leq \mathbf{b}$)

Proposition

Extreme points of conv(X) all lie in X

Hence:

$$\max\{\mathbf{c}^T\mathbf{x} : \mathbf{x} \in X\} \equiv \max\{\mathbf{c}^T\mathbf{x} : \mathbf{x} \in \text{conv}(X)\}$$

However it might require exponential number of inequalities to describe $\text{conv}(X)$
What makes a formulation better than another?

$$X \subseteq \text{conv}(X) \subseteq P_2 \subset P_1$$
$$P_2 \text{ is better than } P_1$$

Definition

Given a set $X \subseteq \mathbb{R}^n$ and two formulations P_1 and P_2 for X, P_2 is a better formulation than P_1 if $P_2 \subset P_1$

Example

$P_1 =$ UFL with $\sum_{i \in M} x_{ij} \le my_j \quad \forall j \in N$
$P_2 =$ UFL with $x_{ij} \le y_j \quad \forall i \in M, j \in N$

$$P_2 \subset P_1$$

- $P_2 \subseteq P_1$ because summing $x_{ij} \le y_j$ over $i \in M$ we obtain $\sum_{i \in M} x_{ij} \le my_j$

- $P_2 \subset P_1$ because there exists a point in P_1 but not in P_2: $m = 6 = 3 \cdot 2 = k \cdot n$

$$x_{10} = 1, x_{20} = 1, x_{30} = 1, \qquad \sum_i x_{i0} \le 6y_0 \;\; y_0 = 1/2$$
$$x_{41} = 1, x_{51} = 1, x_{61} = 1 \qquad \sum_i x_{i1} \le 6y_1 \;\; y_1 = 1/2$$

Resume