# DM545 <br> Linear and Integer Programming 

# Lecture 8 <br> IP Modeling <br> Formulations, Relaxations 

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1. Modeling

Assignment Problem
Set Covering
Graph Problems
Modeling Tricks
2. Formulations

Uncapacited Facility Location
Alternative Formulations
3. Relaxations
4. Well Solved Problems

## Outline

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- Find the cheapest movement for a stacker crane that must pick up and drop objects
- $n$ cities, $c_{i j}$ cost of travel

Variables:

$$
x_{i j}=\left\{\begin{array}{l}
1 \\
0
\end{array}\right.
$$

Objective:

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

## Constraints:

$$
\begin{aligned}
& \sum_{j: j \neq i} x_{i j}=1 \\
& \sum_{i: i \neq j} x_{i j}=1
\end{aligned}
$$

$$
\begin{aligned}
& \forall i=1, \ldots, n \\
& \forall j=1, \ldots, n
\end{aligned}
$$

- cut set constraints

$$
\sum_{i \in S} \sum_{j \neq S} x_{i j} \geq 1
$$

$$
\forall S \subset N, s \neq \emptyset
$$

- subtour elimination constraints

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

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## Modeling Tricks

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

- 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 Tricks I

Minimize the largest of a number of function values:

$$
\min \max \left\{f\left(x_{1}\right), \ldots, f\left(x_{n}\right)\right\}
$$

- Introduce an auxiliary variable $X$ :

$$
\begin{gathered}
\quad X \\
\text { min. } \begin{array}{c}
X\left(x_{1}\right)
\end{array} \leq X \\
f\left(x_{2}\right) \leq X
\end{gathered}
$$

## Modeling Tricks II

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:

$$
a_{1} x+a_{2} y+a_{3} z \leq b\left(d_{1} x+d_{2} y+d_{3} z\right)
$$

which gives:

$$
\left(a_{1}-b d_{1}\right) x+\left(a_{2}-b d_{2}\right) y+\left(a_{3}-b d_{3}\right) z \leq 0
$$

## III "Either/Or Constraints"

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

- $a_{1} x_{1}+a_{2} x_{2} \leq b_{1}$ or
- $d_{1} x_{1}+d_{2} x_{2} \leq b_{2}$

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

- $a_{1} x_{1}+a_{2} x_{2} \leq b_{1}+M y$
- $d_{1} x_{1}+d_{2} x_{2} \leq b_{2}+M(1-y)$


## III "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:

$$
\begin{array}{ll}
A x \leq b+M y & \text { if } y=0 \text { then this is active } \\
A^{\prime} x \leq b^{\prime}+M(1-y) & \text { if } y=1 \text { then this is active }
\end{array}
$$

## IV "Either/Or Constraints"

Generally:

$$
\begin{gathered}
a_{11} x_{1}+a_{12} x_{2}+a_{13} x_{3}+\ldots+a_{1 m} x_{m} \leq d_{1} \\
a_{21} x_{1}+a_{22} x_{2}+a_{23} x_{3}+\ldots+a_{2 m} x_{m} \leq d_{2} \\
\vdots \\
a_{m 1} x_{1}+a_{N 2} x_{2}+a_{N 3} x_{3}+\ldots+a_{N m} x_{m} \leq d_{N}
\end{gathered}
$$

Only $K$ of the $N$ constraints must be satisfied

## IV "Either/Or Constraints"

Introduce binary variables $y_{1}, y_{2}, \ldots, y_{N}$ and a large number $M$

$$
\begin{gathered}
a_{11} x_{1}+a_{12} x_{2}+a_{13} x_{3}+\ldots+a_{1 m} x_{m} \leq d_{1}+M y_{1} \\
a_{21} x_{1}+a_{22} x_{2}+a_{23} x_{3}+\ldots+a_{2 m} x_{m} \leq d_{2}+M y_{2} \\
\vdots \\
a_{m 1} x_{1}+a_{N 2} x_{2}+a_{N 3} x_{3}+\ldots+a_{N m} x_{m} \leq d_{N}+M y_{N} \\
y_{1}+y_{2}+\ldots y_{N}=N-K
\end{gathered}
$$

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

## IV "Either/Or Constraints"

At least $h \leq k$ of $\sum_{j=1}^{n} a_{i j} x_{j} \leq b_{i}, i=1, \ldots, k$ must be satisfied introduce $y_{i}, i=1, \ldots, k$ auxiliary binary variables

$$
\begin{gathered}
\sum_{j=1}^{n} a_{i j} x_{j} \leq b_{i}+M y_{i} \\
\sum_{i} y_{i} \leq k-h
\end{gathered}
$$

## V "Possible Constraints Values"

A constraint must take on one of $N$ given values:

$$
\begin{gathered}
a_{1} x_{1}+a_{2} x_{2}+a_{3} x_{3}+\ldots+a_{m} x_{m}=d_{1} \text { or } \\
a_{1} x_{1}+a_{2} x_{2}+a_{3} x_{3}+\ldots+a_{m} x_{m}=d_{2} \text { or } \\
\vdots \\
a_{1} x_{1}+a_{2} x_{2}+a_{3} x_{3}+\ldots+a_{m} x_{m}=d_{N}
\end{gathered}
$$

Introduce binary variables $y_{1}, y_{2}, \ldots, y_{N}$ :

$$
\begin{gathered}
a_{1} x_{1}+a_{2} x_{2}+a_{3} x_{3}+\ldots+a_{m} x_{m}=d_{1} y_{1}+d_{2} y_{2}+\ldots d_{N} y_{N} \\
y_{1}+y_{2}+\ldots y_{N}=1
\end{gathered}
$$

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Well Solved Problems

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## Uncapacited Facility Location (UFL)

## Given:

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

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

Variables: $y_{j}=\left\{\begin{array}{ll}1 & \text { if depot open } \\ 0 & \text { otherwise }\end{array}, x_{i j}\right.$ fraction of demand of $i$ satisfied by $j$ Objective:

$$
\min \sum_{i \in M} \sum_{j \in N} c_{i j} x_{i j}+\sum_{j \in N} f_{j} y_{j}
$$

## Constraints:

$$
\begin{array}{ll}
\sum_{j=1}^{n} x_{i j}=1 & \forall i=1, \ldots, m \\
\sum_{i \in M} x_{i j} \leq m y_{j} & \forall j \in N
\end{array}
$$

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## 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 $X=P \cap\left(\mathbb{Z}^{n} \times \mathbb{R}^{p}\right)$

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:

$$
\begin{aligned}
\operatorname{conv}(X)= & \left\{x: x=\sum_{i=1}^{t} \lambda_{i} x^{i}, \sum_{i=1}^{t} \lambda_{i}=1, \lambda_{i} \geq 0, \text { for } i=1, \ldots, t,\right. \\
& \text { for all finite subsets } \left.\left\{x^{1}, \ldots, x^{t}\right\} \text { of } X\right\}
\end{aligned}
$$

## Proposition

$\operatorname{conv}(X)$ is a polyhedron

Proposition
Extreme points of $\operatorname{conv}(X)$ all lie in $X$
Hence:

$$
\max \left\{c^{\top} x: x \in X\right\} \equiv \max \left\{c^{\top} x: x \in \operatorname{conv}(x)\right\}
$$

However it might require exponential number of inequalities to describe conv ( $x$ )
What makes a formulation better than another?

$$
\begin{gathered}
X \subseteq \operatorname{conv}(X) \subseteq P_{1} \subset P_{2} \\
P_{1} \text { is better than } P_{2}
\end{gathered}
$$

Definition
Given a set $X \subseteq \mathbb{R}^{n}$ and two formulations $P_{1}$ and $P_{2}$ for $X, P_{1}$ is a better formulation than $P_{2}$ if $P_{1} \subset P_{2}$

Example
$P_{1}=$ UFL with $\sum_{i \in M} x_{i j} \leq m y_{j} \quad \forall j \in N$
$P_{2}=U F L$ with $x_{i j} \leq y_{j} \quad \forall i \in M, j \in N$

$$
P_{2} \subset P_{1}
$$

- $P_{2} \subseteq P_{1}$ because summing $x_{i j} \leq y_{j}$ over $i \in M$ we obtain $\sum_{i \in M} x_{i j} \leq m y_{j}$
- $P_{2} \subset P_{1}$ because there exists a point in $P_{1}$ but not in $P_{2}$ : $m=6=3 \cdot 2=k \cdot n$

$$
\begin{array}{ll}
x_{10}=1 x_{20}=1 x_{30}=1 & \sum_{i} x_{i 0} \leq 6 y_{0} \\
x_{41}=1 x_{51}=1 / 2 \\
\sum_{i} x_{i 1} \leq 6 y_{1} & y_{1}=1 / 2
\end{array}
$$

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## Optimality and Relaxation

$$
z=\max \left\{c(x): x \in X \subseteq \mathbb{Z}^{n}\right\}
$$

How can we prove that $x^{*}$ is optimal? $\bar{z}$ UB
$\underline{z}$ LB
stop when $\bar{z}-\underline{z} \leq \epsilon$


- Primal bounds (here lower bounds): every feasible solution gives a LP may be easy or hard, heuristics
- Dual bounds (here upper bounds): Relaxations

Proposition

$$
\begin{aligned}
(R P) z^{R} & =\max \left\{f(x): x \in T \subseteq \mathbb{R}^{n}\right\} \text { is a relaxation of } \\
(I P) z & =\max \left\{c(x): x \in X \subseteq \mathbb{R}^{n}\right\} \text { if: }
\end{aligned}
$$

(i) $x \subseteq T$ or
(ii) $f(x) \geq c(x) \forall x \in X$

## Relaxations

How to construct relaxations?

1. IP : $\max \left\{c^{T} x: x \in P \cap \mathbb{Z}^{n}\right\}, P=\left\{c \in \mathbb{R}^{n}: A x \leq b\right\}$
$L P: \max \left\{c^{\top} x: x \in P\right\}$
Better formulations give better bounds ( $P_{1} \subseteq P_{2}$ )

Proposition
(i) If a relaxation $R P$ is infeasible, the original problem $O P$ is infeasible.
(ii) Let $x^{*}$ optimal solution for $R P$. If $x^{*} \in X$ and $f\left(x^{*}\right)=c\left(x^{*}\right)$ then $x^{*}$ is optimal for IP.
2. Combinatorial relaxations to easy problems that can be solved rapidly Eg: TSP to Assignment problem Eg: Symmetric TSP to 1-tree
3. Lagrangian relaxation

$$
\begin{aligned}
& I P: \quad z=\max \left\{c^{T} x: A x \leq b, x \in X \subseteq \mathbb{Z}^{n}\right\} \\
& z(u)=\max \left\{c^{T} x+u(b-A x): x \in X\right\} \\
& z(u) \geq z \quad \forall u \geq 0
\end{aligned}
$$

4. Duality:

## Definition

Two problems:

$$
z=\max \{c(x): x \in X\} \quad w=\min \{w(u): u \in U\}
$$

for a weak-dual pair if $c(x) \leq w(u)$ for all $x \in X$ and all $u \in U$. When $z=w$ they form a strong-dual pair

Proposition
$z=\max \left\{c^{\top} x: A x \leq b, x \in \mathbb{Z}_{+}^{n}\right\}$ and $w^{L P}=\min \left\{u b^{T}: u A \geq c, u \in \mathbb{R}_{+}^{m}\right\}$ (ie, linear relaxations) form a weak-dual pair.

## Proposition

Let IP and D be weak-dual pair:
(i) If $D$ us unbounded, then IP is infeasible
(ii) If $x^{*} \in X$ and $u^{*} \in U$ satisfy $c\left(x^{*}\right)=w\left(u^{*}\right)$ then $x^{*}$ is optimal for IP and $u^{*}$ is optimal for $D$.

The advantage is that we do not need to solve an LP like in the LP relaxation to have a bound, any feasible dual solution gives a bound.

## Examples

Weak pairs:
Matching: $\quad z=\max \left\{1^{T} x: A x \leq 1, x \in \mathbb{Z}_{+}^{n}\right\}$
V. Covering: $\quad w=\min \left\{1^{\top} x: A x \geq 1, x \in \mathbb{Z}_{+}^{n}\right\}$

Proof: consider LP relaxations, then $z \leq z^{L P}=w^{L P} \leq w$. (strong when graphs are bipartite)

Weak pairs:
Packing:

$$
z=\max \left\{1^{T} x: A x \leq 1, x \in \mathbb{Z}_{+}^{n}\right\}
$$

S. Covering: $\quad w=\min \left\{1^{T} x: A x \geq 1, x \in \mathbb{Z}_{+}^{n}\right\}$

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## Separation problem

$\max \left\{c^{\top} x: x \in X\right\} \equiv \max \left\{c^{\top} x: x \in \operatorname{conv}(x)\right\}$
$X \subseteq \mathbb{Z}^{n}, P$ a polyhedron $P \subseteq \mathbb{R}^{n}$ and $X=P \cap \mathbb{Z}^{n}$
Definition (Separation problem for a COP)
Given $x^{*} \in P$ is $x^{*} \in \operatorname{conv}(X)$ ? If not find an inequality $a x \leq b$ satisfied by all points in $X$ but violated by the point $x^{*}$.
(Farkas lemma states the existence of such an inequality.)

## Properties of Easy Problems

Four properties that often go together:
Definition
(i) Efficient optimization property: $\exists$ a polynomial algorithm for $\max \left\{c x: x \in X \subseteq \mathbb{R}^{n}\right\}$
(ii) Strong duality property: $\exists$ strong dual $\mathrm{D} \min \{w(u): u \in U\}$ that allows to quickly verify optimality
(iii) Efficient separation problem: $\exists$ efficient algorithm for separation problem
(iv) Efficient convex hull property: a compact description of the convex hull is available

Example:
If explicit convex hull strong duality holds efficient separation property (just description of $\operatorname{conv}(X))$

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