DM545 Linear and Integer Programming

Lecture 8 IP Modeling Formulations, Relaxations

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Modeling Formulations Relaxations Well Solved Problems

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Assignment Problem Set Covering Graph Problems Modeling Tricks

2. Formulations

Uncapacited Facility Location Alternative Formulations

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Traveling Salesman Problem

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

Variables:

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

Objective:

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

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Constraints:

 $\sum_{\substack{j:j \neq i \ j \neq i}} x_{ij} = 1$ $orall i = 1, \dots, n$ $\sum_{\substack{i:i \neq j \ j = 1}} x_{ij} = 1$ $orall j = 1, \dots, n$

cut set constraints



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

subtour elimination constraints

$$\sum_{i\in S}\sum_{j\in S}x_{ij}\leq |S|-1$$

 $\forall S \subset N, 2 \leq |S| \leq n-1$

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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 $\max\{f(x_1),\ldots,f(x_n)\}$

 ► Introduce an auxiliary variable X: min X s. t. f(x₁) ≤ X f(x₂) ≤ X

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Modeling Tricks II

Constraints include variable division:

Constraint of the form

 $\frac{a_1x + a_2y + a_3z}{d_1x + d_2y + d_3z} \le b$

► Rearrange:

 $a_1x + a_2y + a_3z \le b(d_1x + d_2y + d_3z)$

which gives:

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

III "Either/Or Constraints"

In conventional mathematical models, the solution must satisfy all constraints.

Suppose that your constraints are "either/or":

- $a_1x_1 + a_2x_2 \le b_1$ or
- $\bullet \ d_1x_1 + d_2x_2 \le b_2$

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

- $a_1x_1 + a_2x_2 \le b_1 + My$
- $d_1x_1 + d_2x_2 \le b_2 + M(1-y)$

III "Either/Or Constraints"

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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:

 $Ax \le b + My$ if y = 0 then this is active $A'x \le b' + M(1 - y)$ if y = 1 then this is active

IV "Either/Or Constraints"

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Generally:

$$a_{11}x_1 + a_{12}x_2 + a_{13}x_3 + \ldots + a_{1m}x_m \le d_1$$

$$a_{21}x_1 + a_{22}x_2 + a_{23}x_3 + \ldots + a_{2m}x_m \le d_2$$

$$\vdots$$

$$a_{m1}x_1 + a_{N2}x_2 + a_{N3}x_3 + \ldots + a_{Nm}x_m \le d_N$$

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

$$a_{11}x_{1} + a_{12}x_{2} + a_{13}x_{3} + \dots + a_{1m}x_{m} \le d_{1} + My_{1}$$

$$a_{21}x_{1} + a_{22}x_{2} + a_{23}x_{3} + \dots + a_{2m}x_{m} \le d_{2} + My_{2}$$

$$\vdots$$

$$a_{m1}x_{1} + a_{N2}x_{2} + a_{N3}x_{3} + \dots + a_{Nm}x_{m} \le d_{N} + My_{N}$$

$$y_{1} + y_{2} + \dots + y_{N} = N - K$$

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

IV "Either/Or Constraints"

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At least $h \le k$ of $\sum_{j=1}^{n} a_{ij}x_j \le b_i$, i = 1, ..., k must be satisfied introduce y_i , i = 1, ..., k auxiliary binary variables

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

V "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 :

$$a_1x_1 + a_2x_2 + a_3x_3 + \ldots + a_mx_m = d_1y_1 + d_2y_2 + \ldots d_Ny_N$$

 $y_1 + y_2 + \ldots y_N = 1$

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Uncapacited Facility Location (UFL)

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Given:

- depot, $N = \{1, ..., n\}$
- 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 to serves which client

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

Objective:

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

Constraints:

$$\begin{split} \sum_{j=1}^{n} x_{ij} &= 1 & \forall i = 1, \dots, m \\ \sum_{i \in M} x_{ij} &\leq m y_j & \forall j \in N \end{split}$$

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

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:

$$\operatorname{conv}(X) = \left\{ x : x = \sum_{i=1}^{t} \lambda_i x^i, \sum_{i=1}^{t} \lambda_i = 1, \lambda_i \ge 0, \text{ for } i = 1, \dots, t, \right.$$
for all finite subsets $\{x^1, \dots, x^t\}$ of X

Proposition conv(X) is a polyhedron

Proposition

Extreme points of conv(X) all lie in X

Hence:

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

However it might require exponential number of inequalities to describe conv(x)

What makes a formulation better than another?

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

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

$$\begin{array}{l} P_1 = \mathsf{UFL} \text{ with } \sum_{i \in M} x_{ij} \leq m y_j \quad \forall j \in N \\ P_2 = \mathsf{UFL} \text{ with } x_{ij} \leq y_j \quad \forall i \in M, j \in N \end{array}$$

 $P_2 \subset P_1$

- ▶ $P_2 \subseteq P_1$ because summing $x_{ij} \leq y_j$ over $i \in M$ we obtain $\sum_{i \in M} x_{ij} \leq my_j$
- P₂ ⊂ P₁ because there exists a point in P₁ but not in P₂: m = 6 = 3 · 2 = k · n
 - $\begin{array}{ll} x_{10} = 1 \ x_{20} = 1 \ x_{30} = 1 \\ x_{41} = 1 \ x_{51} = 1 \ x_{61} = 1 \end{array} \qquad \qquad \qquad \sum_{i} x_{i0} \le 6y_0 \ y_0 = 1/2 \\ \sum_{i} x_{i1} \le 6y_1 \ y_1 = 1/2 \end{array}$

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Optimality and Relaxation



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z = \max\{c(x) : x \in X \subseteq \mathbb{Z}^n\}
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How can we prove that x^* is optimal?

\overline{z} \text{ UB}

\underline{z} \text{ LB}

stop when \overline{z} - \underline{z} \leq \epsilon
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- Primal bounds (here lower bounds): every feasible solution gives a LP may be easy or hard, heuristics
- Dual bounds (here upper bounds): Relaxations

Proposition

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(RP) z^R = \max\{f(x) : x \in T \subseteq \mathbb{R}^n\} is a relaxation of
(IP) z = \max\{c(x) : x \in X \subseteq \mathbb{R}^n\} if :
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- (i) $x \subseteq T$ or
- (ii) $f(x) \ge c(x) \ \forall x \in X$

Relaxations

How to construct relaxations?

1. $IP : \max\{c^T x : x \in P \cap \mathbb{Z}^n\}, P = \{c \in \mathbb{R}^n : Ax \le b\}$ $LP : \max\{c^T x : x \in P\}$ Better formulations give better bounds $(P_1 \subseteq P_2)$

Proposition

- (i) If a relaxation RP is infeasible, the original problem OP is infeasible.
- (ii) Let x^* optimal solution for RP. If $x^* \in X$ and $f(x^*) = c(x^*)$ 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

$$IP: \quad z = \max\{c^T x : Ax \le b, x \in X \subseteq \mathbb{Z}^n\} \\ z(u) = \max\{c^T x + u(b - Ax) : x \in X\}$$

 $z(u) \geq z \qquad \forall u \geq 0$

4. Duality:

Definition

Two problems:

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

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

Proposition

 $z = \max\{c^T x : Ax \le b, x \in \mathbb{Z}_+^n\}$ and $w^{LP} = \min\{ub^T : uA \ge c, u \in \mathbb{R}_+^m\}$ (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 $c(x^*) = w(u^*)$ 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: $z = \max\{1^T x : Ax \le 1, x \in \mathbb{Z}_+^n\}$ V. Covering: $w = \min\{1^T x : Ax \ge 1, x \in \mathbb{Z}_+^n\}$

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

Weak pairs:Packing: $z = \max\{1^T x : Ax \le 1, x \in \mathbb{Z}_+^n\}$ S. Covering: $w = \min\{1^T x : Ax \ge 1, x \in \mathbb{Z}_+^n\}$

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Separation problem

 $\max\{c^T x : x \in X\} \equiv \max\{c^T x : x \in \operatorname{conv}(x)\}$ $X \subseteq \mathbb{Z}^n$, 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 $x^* \in conv(X)$? If not find an inequality $ax \leq b$ 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: ∃ a polynomial algorithm for max{cx : x ∈ X ⊆ ℝⁿ}
- (ii) Strong duality property: \exists strong dual 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 conv(X)) Resume

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