DM554/DM545 Linear and Integer Programming

Lecture 10 IP Modeling Formulations, Relaxations

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# 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

$$\begin{array}{ll} \max & \sum\limits_{v \in V} w_e x_e \\ & \sum\limits_{e \in E: v \in e} x_e \leq 1 \quad \forall v \in V \\ & x_e \in \{0,1\} \; \forall e \in E \end{array}$$

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.

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

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

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

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

#### 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

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

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<sub>ij</sub>* 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:**

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

• cut set constraints

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

• subtour elimination constraints

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

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

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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 Tricks I

Minimize the largest of a number of function values:

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

• Introduce an auxiliary variable z:

 $\begin{array}{l} \min \quad z \\ \text{s. t. } f(x_1) \leq z \\ \quad f(x_2) \leq z \end{array}$ 

## 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":

 $a_1x_1+a_2x_2\leq b_1$  or  $d_1x_1+d_2x_2\leq b_2$ 

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

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

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:

 $\begin{array}{ll} Ax \leq b + My & \mbox{if } y = 0 \mbox{ then this is active} \\ A'x \leq b' + M(1-y) & \mbox{if } y = 1 \mbox{ then this is active} \end{array}$ 

# IV "Either/Or Constraints"

Generally:

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

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

$$\vdots$$
  

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

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

$$a_{11}x_1 + a_{12}x_2 + a_{13}x_3 + \ldots + a_{1m}x_m \le d_1 + My_1$$
  

$$a_{21}x_1 + a_{22}x_2 + a_{23}x_3 + \ldots + a_{2m}x_m \le d_2 + My_2$$
  

$$\vdots$$
  

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

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

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

### IV "Either/Or Constraints"

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)

Modeling Formulations

#### Given:

- depots  $N = \{1, \ldots, n\}$
- clients  $M = \{1, \ldots, m\}$
- f<sub>j</sub> fixed cost to use depot j
- transport cost for all orders c<sub>ij</sub>

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

 $\sum_{j=1}^{n} x_{ij} = 1 \qquad \qquad \forall i = 1, \dots, m$  $\sum_{i \in M} x_{ij} \le m y_j \qquad \qquad \forall j \in N$ 

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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:

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

#### Proposition

conv(X) is a polyhedron (ie, representable as  $Ax \leq b$ )

#### Proposition

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Extreme points of conv(X) all lie in X
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Hence:

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

However it might require exponential number of inequalities to describe  $\operatorname{conv}(X)$ 

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

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

 $P_2 \subset P_1$ 

- $P_2 \subseteq P_1$  because summing  $x_{ij} \leq y_j$  over  $i \in M$  we obtain  $\sum_{i \in M} x_{ij} \leq my_j$
- P<sub>2</sub> ⊂ P<sub>1</sub> because there exists a point in P<sub>1</sub> but not in P<sub>2</sub>: m = 6 = 3 · 2 = k · n

$$\begin{aligned} x_{10} &= 1, \, x_{20} = 1, \, x_{30} = 1, \\ x_{41} &= 1, \, x_{51} = 1, \, x_{61} = 1 \end{aligned} \qquad \qquad \begin{aligned} \sum_i x_{i0} &\leq 6y_0 \ y_0 = 1/2 \\ \sum_i x_{i1} &\leq 6y_1 \ y_1 = 1/2 \end{aligned}$$

### Resume

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