

# 15.083J/6.859J Integer Optimization

Lecture 7: Ideal formulations III

# 1 Outline

SLIDE 1

- Minimal counterexample
- Lift and project

# 2 Matching polyhedron

SLIDE 2

$$P_{\text{matching}} = \left\{ \mathbf{x} \mid \begin{array}{l} \sum_{e \in \delta(\{i\})} x_e = 1, \quad i \in V, \\ \sum_{e \in \delta(S)} x_e \geq 1, \quad S \subset V, |S| \text{ odd}, |S| \geq 3, \\ 0 \leq x_e \leq 1, \quad e \in E \end{array} \right\}.$$

- $F$  set of perfect matchings in  $G$ .
- Theorem: For the perfect matching problem

$$P_{\text{matching}} = \text{conv}(F).$$

## 2.1 Proof Outline

SLIDE 3

- $\text{conv}(F) \subset P_{\text{matching}}$ .
- For reverse: Assume  $G = (V, E)$  is a graph such that  $P_{\text{matching}} \not\subset \text{conv}(F)$ , and  $|V| + |E|$  is the smallest.
- $\mathbf{x}$  be an extreme point of  $P_{\text{matching}}$  not in  $\text{conv}(F)$ .
- For each edge  $e = \{u, v\}$ ,  $x_e > 0$ , otherwise we could delete  $e$  from  $E$ .
- $x_e < 1$ , otherwise we could replace  $V$  by  $V \setminus \{u, v\}$  and  $E$  by all edges in  $E$  incident to  $V \setminus \{u, v\}$ .
- $|E| > |V|$ ; otherwise, either  $G$  is disconnected (in this case one of the components of  $G$  will be a smaller counterexample), or  $G$  has a node of degree one (in this case the edge  $e$  incident to  $v$  satisfies  $x_e = 1$ ), or  $G$  is the disjoint union of cycles (in this case the theorem holds trivially).
- $\mathbf{x}$  extreme point of  $P_{\text{matching}}$ , there are  $|E|$  linearly independent tight constraint.
- There exists a  $S \subset V$  with  $|S|$  odd,  $|S| \geq 3$ ,  $|V \setminus S| \geq 3$ , and

$$\sum_{e \in \delta(S)} x_e = 1.$$

- Contract  $V \setminus S$  to a single new node  $u$ , to obtain  $G' = (S \cup \{u\}, E')$ .
- $x'_e = x_e$  for all  $e \in E(S)$ , and for  $v \in S$ ,

$$x'_{\{u,v\}} = \sum_{\{j \in V \setminus S, \{v,j\} \in E\}} x_{\{v,j\}}.$$

$\mathbf{x}'$  satisfies constraints with respect to  $G'$ .

- As  $G$  is a smallest counterexample,  $\mathbf{x}'$  belongs to the convex hull of matchings on  $G'$ ,

$$\mathbf{x}' = \sum_{M'} \lambda_{M'} \chi^{M'}.$$

- Contract  $S$  to a single new node  $t$  we obtain a graph  $G'' = ((V \setminus S) \cup \{t\}, E'')$  and a vector  $\mathbf{x}''$ :

$$\mathbf{x}'' = \sum_{M''} \mu_{M''} \chi^{M''}.$$

- “Glue together” perfect matchings  $M'$  and  $M''$

$$\mathbf{x} = \sum_{e \in \delta(S)} \sum_{M \text{ perfect matching: } M \cap \delta(S) = \{e\}} \frac{\lambda_{M'} \mu_{M''}}{x_e} \chi^M$$

### 3 Lift and project

SLIDE 4

- $S = \{\mathbf{x} \in \mathcal{Z}^n \mid \mathbf{A}\mathbf{x} \leq \mathbf{b}\}$ .
- **(Lift)** Multiply  $\mathbf{A}\mathbf{x} \leq \mathbf{b}$  by  $x_j$  and  $1 - x_j$

$$\begin{aligned} (\mathbf{A}\mathbf{x})x_j &\leq \mathbf{b}x_j & (*) \\ (\mathbf{A}\mathbf{x})(1 - x_j) &\leq \mathbf{b}(1 - x_j) \end{aligned}$$

and substitute  $y_{ij} = x_i x_j$  for  $i, j = 1, \dots, n$ ,  $i \neq j$  and  $x_j = x_j^2$ . Let  $L_j(P)$  be the resulting polyhedron.

- **(Project)** Project  $L_j(P)$  back to the  $\mathbf{x}$  variables by eliminating variables  $\mathbf{y}$ . Let  $P_j$  be the resulting polyhedron, i.e.,  $P_j = (L_j(P))_x$ .

#### 3.1 Theorem

SLIDE 5

$$P_j = \text{conv}(P \cap \{\mathbf{x} \in \mathcal{R}^n \mid x_j \in \{0, 1\}\})$$

Proof:

- $\mathbf{x}' \in P \cap \{\mathbf{x} \in \mathcal{R}^n \mid x_j \in \{0, 1\}\}$  and  $y'_{ij} = x'_i x'_j$ .
- Since  $x'_j = (x'_j)^2$  and  $\mathbf{A}\mathbf{x}' \leq \mathbf{b}$ ,  $(\mathbf{x}', \mathbf{y}') \in L_j(P)$  and thus  $\mathbf{x}' \in P_j$ . Hence,

$$\text{conv}(P \cap \{\mathbf{x} \in \mathcal{R}^n \mid x_j \in \{0, 1\}\}) \subseteq P_j.$$

- If  $P \cap \{\mathbf{x} \in \mathcal{R}^n \mid x_j = 0\} = \emptyset$ , then from the Farkas lemma there exists  $\mathbf{u} \geq \mathbf{0}$ , such that  $\mathbf{u}'\mathbf{A} = -\mathbf{e}_j$  and  $\mathbf{u}'\mathbf{b} = -1$ . Thus, for all  $\mathbf{x}$  satisfying (\*) we have

$$\mathbf{u}'\mathbf{A}\mathbf{x}(1 - x_j) \leq \mathbf{u}'\mathbf{b}(1 - x_j).$$

Hence, for all  $\mathbf{x} \in P_j$

$$-\mathbf{e}'_j \mathbf{x}(1 - x_j) = -x_j(1 - x_j) \leq -(1 - x_j).$$

Replacing  $x_j^2$  by  $x_j$ , we obtain that  $x_j \geq 1$  is valid for  $P_j$ . Since, in addition,  $P_j \subseteq P$ , we conclude that

$$P_j \subseteq P \cap \{\mathbf{x} \in \mathcal{R}^n \mid x_j = 1\} = \text{conv}(P \cap \{\mathbf{x} \in \mathcal{R}^n \mid x_j \in \{0, 1\}\}).$$

- Similarly, if  $P \cap \{\mathbf{x} \in \mathcal{R}^n \mid x_j = 1\} = \emptyset$ , then

$$P_j \subseteq \text{conv}(P \cap \{\mathbf{x} \in \mathcal{R}^n \mid x_j \in \{0, 1\}\}).$$

- Suppose  $P \cap \{\mathbf{x} \in \mathcal{R}^n \mid x_j = 0\} \neq \emptyset$ ,  $P \cap \{\mathbf{x} \in \mathcal{R}^n \mid x_j = 1\} \neq \emptyset$ .
- We prove that all valid inequalities for  $\text{conv}(P \cap \{\mathbf{x} \in \mathcal{R}^n \mid x_j \in \{0, 1\}\})$  are also valid for  $P_j$ .
- $\mathbf{a}'\mathbf{x} \leq \alpha$  a valid inequality for  $\text{conv}(P \cap \{\mathbf{x} \in \mathcal{R}^n \mid x_j \in \{0, 1\}\})$ .
- $\mathbf{x} \in P$ . If  $x_j = 0$ , then for all  $\lambda \in \mathcal{R}$   $\mathbf{a}'\mathbf{x} + \lambda x_j = \mathbf{a}'\mathbf{x} \leq \alpha$ .
- If  $x_j > 0$ , then there exists  $\lambda \leq 0$ , such that for all  $\mathbf{x} \in P$ ,

$$\mathbf{a}'\mathbf{x} + \lambda x_j \leq \alpha.$$

- Analogously, since  $\mathbf{a}'\mathbf{x} \leq \alpha$  is valid for  $P \cap \{\mathbf{x} \in \mathcal{R}^n \mid x_j = 1\}$ , there exists some  $\nu \leq 0$  such that for all  $\mathbf{x} \in P$ ,

$$\mathbf{a}'\mathbf{x} + \nu(1 - x_j) \leq \alpha.$$

- For all  $\mathbf{x}$  satisfying (\*),

$$\begin{aligned} (1 - x_j)(\mathbf{a}'\mathbf{x} + \lambda x_j) &\leq (1 - x_j)\alpha \\ x_j(\mathbf{a}'\mathbf{x} + \nu(1 - x_j)) &\leq x_j\alpha. \end{aligned}$$

- Hence,

$$\mathbf{a}'\mathbf{x} + (\lambda + \nu)(x_j - x_j^2) \leq \alpha.$$

- After setting  $x_j^2 = x_j$  we obtain that for all  $\mathbf{x} \in P_j$ ,  $\mathbf{a}'\mathbf{x} \leq \alpha$ , thus all valid inequalities for  $\text{conv}(P \cap \{\mathbf{x} \in \mathcal{R}^n \mid x_j \in \{0, 1\}\})$  are also valid for  $P_j$ , and thus  $P_j \subseteq \text{conv}(P \cap \{\mathbf{x} \in \mathcal{R}^n \mid x_j \in \{0, 1\}\})$ .

### 3.2 Example

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$$P = \{(x_1, x_2)' \mid 2x_1 - x_2 \geq 0, 2x_1 + x_2 \leq 2, x_1 \geq 0, x_2 \geq 0\}.$$

$$\begin{aligned} 2x_1^2 - x_1x_2 &\geq 0 \\ 2x_1(1 - x_1) - x_2(1 - x_1) &\geq 0 \\ 2x_1^2 + x_1x_2 &\leq 2x_1 \\ 2x_1(1 - x_1) + x_2(1 - x_1) &\leq 2(1 - x_1) \\ x_1^2 &\geq 0 \\ x_1(1 - x_1) &\geq 0 \\ x_2x_1 &\geq 0 \\ x_2(1 - x_1) &\geq 0. \end{aligned}$$

$$y = x_1x_2, x_1^2 = x_1$$

$$\begin{aligned} 2x_1 - y &\geq 0 \\ -x_2 + y &\geq 0 \\ y &\leq 0 \\ x_2 - y &\leq 2 - 2x_1 \\ x_1 &\geq 0 \\ 0 &\geq 0 \\ y &\geq 0 \\ x_2 - y &\geq 0. \end{aligned}$$

This implies that  $y = 0$ ,

$$\begin{aligned} x_1 &\geq 0 \\ -x_2 &\geq 0 \\ x_2 &\leq 2 - 2x_1 \\ x_1 &\geq 0 \\ x_2 &\geq 0, \end{aligned}$$

which leads to

$$\begin{aligned} P_1 &= \{(x_1, x_2)' \mid 0 \leq x_1 \leq 1, x_2 = 0\} \\ &= \text{conv}(P \cap \{(x_1, x_2)' \mid x_1 \in \{0, 1\}\}). \end{aligned}$$

### 3.3 Convex hull

SLIDE 7

- $P_{i_1, i_2, \dots, i_t} = ((P_{i_1})_{i_2} \dots)_{i_t}$ .
- Theorem: The polyhedron  $P_{i_1, i_2, \dots, i_t}$  satisfies:

$$P_{i_1, \dots, i_t} = \text{conv}(P \cap \{\mathbf{x} \in \mathcal{R}^n \mid x_i \in \{0, 1\}, i \in \{i_1, \dots, i_t\}\}).$$

- $P_{1, \dots, n} = P_I$ .

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15.083J / 6.859J Integer Programming and Combinatorial Optimization  
Fall 2009

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