15.083J Integer Programming and Combinatorial Optimization

#### Fall 2009

# Mixed-Integer Programming I

### Mixed-Integer Linear Programming

$$\begin{array}{ll} \max & cx + hy \\ \text{s.t.} & Ax + Gy \leq b \\ & x \text{ integral} \end{array}$$

where c, h, A, G, and b are rational vectors and matrices, respectively.

#### Projections

- Let  $P \subseteq \mathbb{R}^{n+p}$ , where  $(x, y) \in P$  is interpreted as  $x \in \mathbb{R}^n$  and  $y \in \mathbb{R}^p$ .
- The projection of P onto the x-space  $\mathbb{R}^n$  is

$$\operatorname{proj}_{x}(P) = \{ x \in \mathbb{R}^{n} : \exists y \in \mathbb{R}^{p} \text{ with } (x, y) \in P \}.$$

**Theorem 1.** Let  $P = \{(x, y) : Ax + Gy \le b\}$ . Then

$$\operatorname{proj}_{x}(P) = \{ x : v^{t}(b - Ax) \ge 0 \text{ for all } t \in T \},\$$

where  $\{v^t\}_{t\in T}$  is the set of extreme rays of  $\{v : vG = 0, v \ge 0\}$ .

#### The Fundamental Theorem of MILP

**Theorem 2** (Meyer 1974). Given rational matrices A and G and a rational vector b, let  $P = \{(x, y) : Ax + Gy \leq b\}$  and  $S = \{(x, y) \in P : x \text{ integral}\}$ . There exist rational matrices A', G', and a rational vector b' such that

$$\operatorname{conv}(S) = \{(x, y) : A'x + G'y \le b'\}.$$

Proof:

- We may assume that  $S \neq \emptyset$ .
- By the Minkowski-Weyl Theorem,  $P = \operatorname{conv}(V) + \operatorname{cone}(R)$ , where  $V = (v^1, \ldots, v^p)$  and  $R = (r^1, \ldots, r^q)$ .
- We may assume that V is a rational matrix and R is an integral matrix.

• Consider the following truncation of *P*:

$$T = \{(x, y) : (x, y) = \sum_{i=1}^{p} \lambda_i v^i + \sum_{j=1}^{q} \mu_j r^j, \sum_{i=1}^{p} \lambda_i = 1, \\ \lambda \ge 0, 0 \le \mu \le 1\}.$$

- T is bounded and is the projection of a rational polyhedron. It therefore is a rational polytope.
- Let  $T_I = \{(x, y) \in T : x \text{ integral}\}$ . Claim:  $\operatorname{conv}(T_I)$  is a rational polytope.
- Since T is a polytope,  $X = \{x : \exists y \text{ s.th. } (x, y) \in T_I\}$  is finite.
- For fixed  $\bar{x} \in X$ ,  $T_{\bar{x}} = \{(\bar{x}, y) : (\bar{x}, y) \in T_I\}$  is a rational polytope. Hence,  $T_{\bar{x}} = \operatorname{conv}(V_{\bar{x}})$  for some rational matrix  $V_{\bar{x}}$ .
- Since X is finite, there is a rational matrix  $V_{T_I}$  which contains all the columns of all matrices  $V_{\bar{x}}$ , for  $\bar{x} \in X$ .
- Therefore,  $\operatorname{conv}(T_I) = \operatorname{conv}(V_{T_I})$ , which proves the claim.
- $(\bar{x}, \bar{y}) \in S$  iff  $\bar{x}$  is integral and there exist  $\lambda \ge 0$ ,  $\sum_{i=1}^{p} \lambda_i = 1$ , and  $\mu \ge 0$  such that

$$(\bar{x},\bar{y}) = \sum_{i=1}^p \lambda_i v^i + \sum_{j=1}^q (\mu_j - \lfloor \mu_j \rfloor) r^j + \sum_{j=1}^q \lfloor \mu_j \rfloor r^j.$$

- The point  $\sum_{i=1}^{p} \lambda_i v^i + \sum_{j=1}^{q} (\mu_j \lfloor \mu_j \rfloor) r^j$  belongs to T.
- Since  $\bar{x}$  and  $\lfloor \mu_j \rfloor r^j$  are integral it also belongs to  $T_I$ .
- Thus

$$S = T_I + R_I,\tag{1}$$

where  $R_I$  is the set of integral conic combinations of  $r^1, \ldots, r^q$ .

• (1) implies that

$$\operatorname{conv}(S) = \operatorname{conv}(T_I) + \operatorname{cone}(R).$$

- By the above claim  $conv(T_I)$  is a rational polytope.
- Thus  $\operatorname{conv}(S)$  is a rational polyhedron (having the same recession cone as P).

## Union of Polyhedra

- Consider k polyhedra  $P_i = \{x \in \mathbb{R}^n : A_i x \leq b^i\}, i = 1, \dots, k.$
- One can show that  $\overline{\operatorname{conv}}(\cup_{i=1}^k P_i)$  is a polyhedron.

- Furthermore, we will show that this polyhedron can be obtained as the projection onto  $\mathbb{R}^n$  of a polyhedron with polynomially many variables and constraints in a higher-dimensional space.
- (The closure is needed: let  $P_1$  be a single point and let  $P_2$  be a line that does not contain  $P_2$ .)

**Theorem 3.** For i = 1, ..., k, let  $P_i = Q_i + C_i$  be nonempty polyhedra. Then  $Q = \operatorname{conv}(\cup_{i=1}^k Q_i)$  is a polytope,  $C = \operatorname{conv}(\cup_{i=1}^k C_i)$  is a finitely generated cone, and  $\overline{\operatorname{conv}}(\cup_{i=1}^k P_i) = Q + C$ .

- No proof here, but note that the claims on Q and C are straightforward to check.
- One consequence of the proof is that if  $P_1, \ldots, P_k$  have identical recession cones, then  $\operatorname{conv}(\bigcup_{i=1}^k P_i)$  is a polyhedron.

**Theorem 4** (Balas 1974). Consider k polyhedra  $P_i = Q_i + C_i = \{x \in \mathbb{R}^n : A_i x \leq b^i\}$  and let  $Y \subseteq \mathbb{R}^{n+kn+k}$  be the polyhedron defined by

$$A_i x^i \le b^i y_i, \sum_{i=1}^k x^i = x, \sum_{i=1}^k y_i = 1, y_i \ge 0 \text{ for } i = 1, \dots, k$$

Then

$$\operatorname{proj}_x(Y) = Q + C,$$

where  $Q = \operatorname{conv}(\cup_{i=1}^{k}Q_i)$  and  $C = \operatorname{conv}(\cup_{i=1}^{k}C_i)$ .

Proof:

- First, let  $x \in Q + C$ .
- There exist  $w^i \in Q_i$  and  $z^i \in C_i$  such that  $x = \sum_i y_i w^i + \sum_i z^i$ , where  $y_i \ge 0$  and  $\sum_i y_i = 1$ .
- Let  $x^i = y_i w^i + z^i$ . Then  $A_i x^i \le b^i y_i$  and  $x = \sum_i x^i$ .
- This shows  $x \in \operatorname{proj}_x(Y)$ .
- Now, let  $x \in \operatorname{proj}_x(Y)$ .
- There exist  $x^1, \ldots, x^k, y$  such that  $x = \sum_i x^i$  where  $A_i x^i \leq b^i y_i, \sum_i y_i = 1, y \geq 0$ .
- Let  $I = \{i : y_i > 0\}.$
- For  $i \in I$ , let  $z^i = \frac{x^i}{y_i}$ . Then  $z^i \in P_i$ .
- Since  $P_i = Q_i + C_i$ , we can write  $z^i = w^i + \frac{r^i}{y_i}$  where  $w^i \in Q_i$  and  $r^i \in C_i$ .
- For  $i \notin I$ , we have  $A_i x^i \leq 0$ , that is  $x^i \in C_i$ . Let  $r^i = x^i$  for  $i \notin I$ .
- Then,

$$x = \sum_{i \in I} y_i z^i + \sum_{i \notin I} x^i = \sum_{i \in I} y_i w^i + \sum_i r^i \in Q + C$$

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#### Lift-and-Project Revisited

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We consider mixed-0/1 linear programs:

min 
$$cx$$
  
s.t.  $Ax \ge b$   
 $x_j \in \{0,1\}$  for  $j = 1, ..., n$   
 $x_j \ge 0$  for  $j = n+1, ..., n+p$ 

We let  $P = \{x \in \mathbb{R}^{n+p}_+ : Ax \ge b\}$  and  $S = \{x \in \{0,1\}^n \times \mathbb{R}^p_+ : Ax \ge b\}.$ We assume that  $Ax \ge b$  includes  $-x_j \ge -1$  for  $j = 1, \ldots, n$ , but not  $x \ge 0$ .

• Given an index  $j \in \{1, \ldots, n\}$ , let

$$P_j = \operatorname{conv} \{ (Ax \ge b, x \ge 0, x_j = 0) \cup (Ax \ge b, x \ge 0, x_j = 1) \}$$

- By definition, this is the tightest possible relaxation among all relaxations that ignore the integrality of all variables  $x_i, i \neq j$ .
- $\bigcap_{j=1}^{n} P_j$  is called the *lift-and-project closure*:

$$\operatorname{conv}(S) \subseteq \bigcap_{j=1}^{n} P_j \subseteq P.$$

• On 35 mixed-0/1 linear programs from MIPLIB, the lift-and-project closure reduces the integrality gap by 37% on average [Bonami & Minoux 2005].

#### Lift-and-Project Cuts

 $P_j$  is the convex hull of the union of two polyhedra:

$$\begin{array}{ll} Ax \geq b & & Ax \geq b \\ x \geq 0 & & \text{and} & & x \geq 0 \\ -x_j \geq 0 & & & x_j \geq 1 \end{array}$$

By the above theorem:

$$P_{j} = \operatorname{proj}_{x} \begin{pmatrix} Ax^{0} \geq by_{0} \\ -x_{j}^{0} \geq 0 \\ Ax^{1} \geq by_{1} \\ x_{j}^{1} \geq y_{1} \\ x^{0} + x^{1} = x \\ y_{0} + y_{1} = 1 \\ x^{0}, x^{1}, y_{0}, y_{1} \geq 0 \end{pmatrix}$$

• Using the projection theorem, we get that  $P_j$  is defined by the inequalities  $\alpha x \ge \beta$  such that

- Such an inequality  $\alpha x \geq \beta$  is called a lift-and-project inequality.
- Given a fractional point  $\bar{x}$ , we can determine if there exists a lift-and-project inequality  $\alpha x \geq \beta$  valid for  $P_j$  that cuts off  $\bar{x}$ .
- This problem amounts to finding  $(\alpha, \beta, u, u_0, v, v_o)$  satisfying (2) such that  $\alpha \bar{x} \beta < 0$ .
- In order to find a "best" cut in cone (2), we solve the *cut-generating LP*:

#### Mixed Integer Inequalities

- Consider  $S = \{(x, y) \in \mathbb{Z}_+^n \times \mathbb{R}_+^p : \sum_{j=1}^n a_j x_j + \sum_{j=1}^p g_j y_j = b\}.$
- Let  $b = \lfloor b \rfloor + f_0$  where  $0 < f_0 < 1$ .
- Let  $a_j = \lfloor a_j \rfloor + f_j$  where  $0 \le f_j < 1$ .
- Then  $\sum_{f_j \le f_0} f_j x_j + \sum_{f_j > f_0} (f_j 1) x_j + \sum_{j=1}^p g_j y_j = k + f_0$ , where k is some integer.
- Since  $k \leq -1$  or  $k \geq 0$ , any  $x \in S$  satisfies

$$\sum_{f_j \le f_0} \frac{f_j}{f_0} x_j - \sum_{f_j > f_0} \frac{1 - f_j}{f_0} x_j + \sum_{j=1}^p \frac{g_j}{f_0} y_j \ge 1$$
(3)

OR

$$-\sum_{f_j \le f_0} \frac{f_j}{1 - f_0} x_j + \sum_{f_j > f_0} \frac{1 - f_j}{1 - f_0} x_j - \sum_{j=1}^p \frac{g_j}{1 - f_0} y_j \ge 1.$$
(4)

• This is of the form  $\sum_j a_j^1 x_j \ge 1$  or  $\sum_j a_j^2 x_j \ge 1$ , which implies  $\sum_j \max\{a_j^1, a_j^2\} x_j \ge 1$  for any  $x \ge 0$ .

- For each variable, what is the max coefficient in (3) and (4)?
- We get

$$\sum_{f_j \le f_0} \frac{f_j}{f_0} x_j + \sum_{f_j > f_0} \frac{1 - f_j}{1 - f_0} x_j + \sum_{g_j > 0} \frac{g_j}{f_0} y_j - \sum_{g_j < 0} \frac{g_j}{1 - f_0} y_j \ge 1.$$

- This is the Gomory mixed integer (GMI) inequality.
- In the pure integer programming case, the GMI inequality reduces to

$$\sum_{f_j \le f_0} \frac{f_j}{f_0} x_j + \sum_{f_j > f_0} \frac{1 - f_j}{1 - f_0} x_j \ge 1.$$

• Since  $\frac{1-f_j}{1-f_0} < \frac{f_j}{f_0}$  when  $f_j > f_0$ , the GMI inequality dominates

$$\sum_{j=1}^{n} f_j x_j \ge f_0$$

which is known as the *fractional cut*.

- Consider now  $S = \{(x, y) \in \mathbb{Z}^n_+ \times \mathbb{R}^p_+ : Ax + Gy \le b\}.$
- Let  $P = \{(x, y) \in \mathbb{R}^n_+ \times \mathbb{R}^p_+ : Ax + Gy \le b\}$  be the underlying polyhedron.
- Let  $\alpha x + \gamma y \leq \beta$  be any valid for *P*.
- Add a nonnegative slack variable s, use  $\alpha x + \gamma y + s = \beta$  to derive a GMI inequality, and eliminate  $s = \beta \alpha x \gamma y$  from it.
- The result is a valid inequality for S.
- These inequalities are called the GMI inequalities for S.
- In contrast to lift-and-project cuts, it is in general NP-hard to find a GMI inequality that cuts off a point  $(\bar{x}, \bar{y}) \in P \setminus S$ , or show that none exists.
- However, one can easily find a GMI inequality that cuts off a basic feasible solution.
- On 41 MIPLIB instances, adding the GMI cuts generated from the optimal simplex tableaux reduces the integrality gap by 24% on average [Bonami et al. 2008]
- GMI cuts are widely used in commercial codes today.
- Numerical issues need to be addressed, however.

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