Solving Mixed Integer Linear Programs using Cutting Planes

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Mixed Integer Linear Program and Linear Program Relaxation

MILP (standard form)

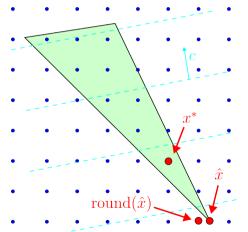
$$\begin{array}{ll}
\underset{x}{\text{min}} & c \mid x \\
\text{s.t.} & x \in F \\
& := \{ x \mid Ax \le b, \ x \in \mathbb{Z}_{\ge 0}^{n_1} \times \mathbb{R}_{\ge 0}^{n-n_1} \}
\end{array}$$

$$c, x \in \mathbb{R}^n, A \in \mathbb{R}^{m \times n}, b \in \mathbb{R}^m$$

LP Relaxation

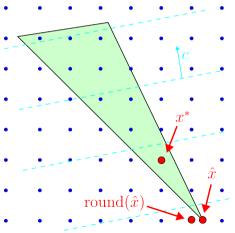
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\end{array}$$

Mixed Integer Linear Program



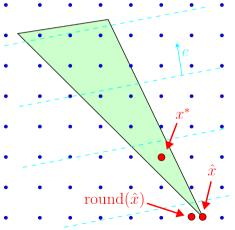
■ In general, round(\hat{x}) $\neq x^*$

Observations



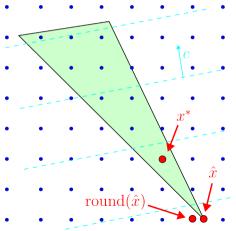
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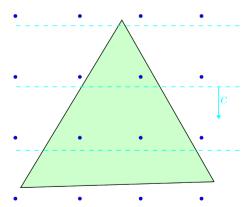


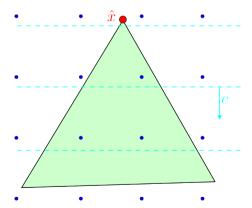
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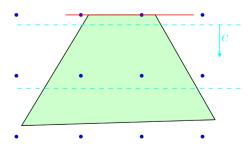


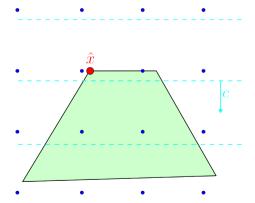
- In general, round(\hat{x}) $\neq x^*$
- The relaxed solution gives a lower bound: $c^{\top}\hat{x} < c^{\top}x^*$.
- \hat{x} can be found at a vertex of P (Simplex Algorithm).
- If \hat{x} is already feasible, then $c^{\top}\hat{x} = c^{\top}x^*$.

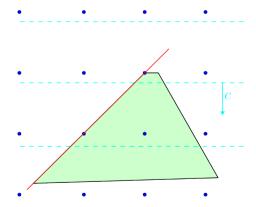


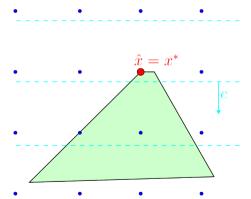












Valid Inequalities and Cuts

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- An inequality $a^{\top}x \leq r$ is <u>valid</u> for a set F if $a^{\top}x \leq r$ is satisfied for all $x \in F$.
 - For example: $x \le 2$ is a valid inequality for $\{x \in \mathbb{Z}_{>0} \mid x \le 2.718\}$
- A <u>cutting plane</u> (or cut) w.r.t. $\hat{x} \in P \setminus F$ is any valid inequality $a^{\top}x \leq r$ for F such that:

$$a^{\mathsf{T}}\hat{x} > r$$

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1: LP \leftarrow Relaxation of the MILP

2: repeat

3: \hat{x} \leftarrow Optimal solution of the LP

4: if (\hat{x}_1, ..., \hat{x}_{n_1}) \notin \mathbb{Z}^{n_1} then

5: Add a cut w.r.t. \hat{x} to the LP

6: until (\hat{x}_1, ..., \hat{x}_{n_1}) \in \mathbb{Z}^{n_1}

7: return \hat{x}
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Cutting Strategy

Question: How to generate "good" and useful cuts?

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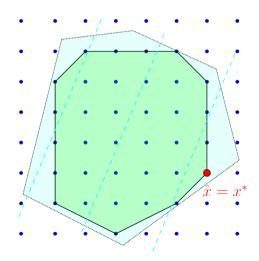
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Question: How to generate "good" and useful cuts?

- Good: Cut away as much as possible (while staying feasible)
- Useful: Cut away the optimal solution of the relaxation

Convex Hull

- The relaxed solution \hat{x} in conv(F) also solves the MILP.
- But computing the convex hull is infeasible (exponential).



Any real number $a \in \mathbb{R}$ can be expressed as

$$a = |a| + f_a$$

for some unique $\lfloor a \rfloor \in \mathbb{Z}$ and $f_a \in [0, 1)$.

- $\lfloor a \rfloor = \max\{z \in \mathbb{Z} \mid z \leq a\}$ is the integer part of a.
- $f_a = a \lfloor a \rfloor$ is the fractional part of a.

Integer Part and Fractional Part

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- $\lfloor -a \rfloor = -\lceil a \rceil$ where $\lceil a \rceil = \min\{z \in \mathbb{Z} \mid z \geq a\}$
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Integer Part and Fractional Part

Any real number $a \in \mathbb{R}$ can be expressed as

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for some unique $|a| \in \mathbb{Z}$ and $f_a \in [0, 1)$.

- $|a| = \max\{z \in \mathbb{Z} \mid z < a\}$ is the integer part of a.
- \bullet $f_a = a |a|$ is the fractional part of a.

$$f_a = 0 \Leftrightarrow a = |a| \Leftrightarrow a \in \mathbb{Z}$$

- |-a| = -[a] where $\lceil a \rceil = \min\{z \in \mathbb{Z} \mid z > a\}$
- $\blacksquare a \in \mathbb{Z}$ and $a < b \Rightarrow a < |b|$
- \blacksquare $a \in \mathbb{Z}$ and $a > b \Rightarrow a > \lceil b \rceil$

Chvátal-Gomory Inequality for Integer Linear Programs

Let $\sum_{j=1}^{n} a_{ij} x_j \leq b_i$ for an Integer Linear Program $(x \in \mathbb{Z}_{\geq 0}^n)$. Then the following inequalities are valid for any $\alpha \geq 0$:

$$\alpha \geq 0$$

$$\sum_{i=1}^{n} \lfloor \alpha a_{ij} \rfloor x_j \leq \alpha b_i$$

$$x_j \ge 0$$

$$\sum_{i=1}^{n} \lfloor \alpha a_{ij} \rfloor x_j \leq \lfloor \alpha b_i \rfloor$$

$$x_i \in \mathbb{Z}$$

Example

Illustration

Chvátal-Gomory Inequality

If not all variables are integer, these inequalities are not valid. Show that it does not work for mixed integer problem

Basic Mixed Integer Rounding Inequalities I

Let
$$x \in \mathbb{Z}_{\geq 0}$$
, $y \in \mathbb{R}_{\geq 0}$, $b \in \mathbb{R}_{\geq 0} \setminus \mathbb{Z}$. Then

$$x \le \lfloor b \rfloor$$
 is a valid inequality for $\{x + y \le b\}$ (1)

and

$$x \ge \lceil b \rceil$$
 is a valid inequality for $\{-x + y \le -b\}$ (2)

Basic Mixed Integer Rounding Inequalities I

Hello World

Basic Mixed Integer Rounding Inequalities II

Let $x \in \mathbb{Z}_{>0}$, $y \in \mathbb{R}_{>0}$, $b \in \mathbb{R}_{>0} \setminus \mathbb{Z}$. Then

$$x - \frac{1}{f_b - 1} \le \lfloor b \rfloor \text{ is a valid inequality for } \{x - y \le b\}$$
 (1)

and

$$x + \frac{1}{f_b} \ge \lceil b \rceil$$
 is a valid inequality for $\{-x - y \le -b\}$ (2)

Basic Mixed Integer Rounding Inequalities II

General Mixed Integer Rounding Inequality

Let $F_{MIR} = \{(x, y) \in \mathbb{Z}^2_{\geq 0} \times \mathbb{R}_{\geq 0} \mid a_1x_1 + a_2x_2 - y \leq b\}$ where $a \in \mathbb{R}^2$, $b \in \mathbb{R} \setminus \mathbb{Z}$ and assume that $f_1 \leq f_b \leq f_2$. Then the inequality

$$\lfloor a_1 \rfloor x_1 + \left(\lfloor a_2 \rfloor + \frac{f_2 - f_b}{1 - f_b} \right) x_2 - \frac{1}{1 - f_b} y \le \lfloor b \rfloor$$

is valid for F_{MIR} .

General Mixed Integer Rounding Inequality

Simplex Algorithm

Simplex finds $\hat{x} \in P \times \mathbb{R}^{N-n}_{>0}$ and creates the optimal simplex tableau:

i—th row in the simplex tableau

$$x_{B_i} + \sum_{j \in NB} \bar{a}_{ij} x_j = \bar{b}_i$$

- $x_1, ..., x_{n_1}$: Integer problem variables
- $x_{n_1+1}, ..., x_n$: Real problem variables
- $x_{n+1}, ..., x_N$: (Real) slack variables

- $B = \{B_1, ..., B_m\}$: Basic variables
- $NB = \{1, ..., N\} \setminus B$: Nonbasic variables $(\hat{x}_j = 0 \text{ for } j \in NB)$

Gomory Mixed Integer Cut

Let $N_1 = NB \cap \{1, ..., n_1\}$, $N_2 = NB \cap \{n_1 + 1, ..., x_N\}$. Consider the i-th row in the optimal simplex tableau

$$x_{B_i} + \sum_{j \in \mathcal{N}_1} \bar{a}_{ij} x_j + \sum_{j \in \mathcal{N}_2} \bar{a}_{ij} x_j = \bar{b}_i$$

and assume $B_i \leq n_1$ but $\hat{x}_{B_i} = \bar{b}_i \notin \mathbb{Z}$. Then the Gomory Mixed Integer Cut

$$x_{B_i} + \sum_{\substack{j \in N_1 \\ f_{ij} \le f_i}} \lfloor \bar{a}_{ij} \rfloor x_j + \sum_{\substack{j \in N_1 \\ f_{ij} > f_i}} \left(\lfloor \bar{a}_{ij} \rfloor + \frac{f_{ij} - f_i}{1 - f_i} \right) x_j + \sum_{\substack{j \in N_2 \\ \bar{a}_{ij} < 0}} \left(\frac{\bar{a}_{ij}}{1 - f_i} \right) x_j \le \lfloor \bar{b}_i \rfloor$$

is a valid inequality for F that is not satisfied by \hat{x} .

Gomory Mixed Integer Cut

- Let a MILP be given with feasible region $F = \{x \in \mathbb{Z}_{\geq 0}^{n_1} \times \mathbb{R}_{\geq 0}^{n_{-n_1}} \mid Ax \leq b\}$ for some $A \in \mathbb{R}^{m \times n}$, $b \in \mathbb{R}^m$.
- The relaxation is the LP obtained by removing the integer constraints, so its feasible region is the polyhedron $P = \{x \in \mathbb{R}^n_{>0} \mid Ax \leq b\}$.
- Repeat the following two steps until $\hat{x} \in F$:
 - **1** Solve the LP using the Simplex Algorithm and obtain $\hat{x} \in P$
 - 2 TODO

Project Demonstration

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 - Heuristic to evaluate the efficiency of a cutting plane (e.g. euclidean distance to \hat{x}).
 - Add multiple cutting planes in each iteration.
- Other cutting plane strategies exist.

Branch & Bound

- Like Cutting Planes, we solve the problem relaxation and add constraints until an optimal solution has been found.
- Divide & Conquer

Branching

Cut off the non-integer neighborhood of $x_i^* \notin \mathbb{Z} \Rightarrow$ Two new relaxation problems

Branch & Bound

Bounding

- Initially, we only know that $-\infty < c^{\top} x_{MHP}^* < \infty$ (not very helpful).
- Improve bounds until Upper Bound Lower Bound $< \epsilon$:
 - Lower Bound: For any LP-Relaxation, we have $c^{\top}x_{LP}^* \leq c^{\top}x_{MILP}^*$
 - Upper Bound: By definition, $c^{\top}x_{MILP}^* \leq c^{\top}x$ for any feasible $x \in F_{MILP}$

Branch & Cut

- Hello World
- Cutting Planes + Branch & Bound = Branch & Cut

Questions?

References