

MA427 Lecture 5
Integer Programming
Branch-and-bound and good formulations

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Today's lecture

- ▶ Branch-and-bound
- ▶ Good formulations
- ▶ Knapsack cover inequalities
- ▶ Ideal formulations

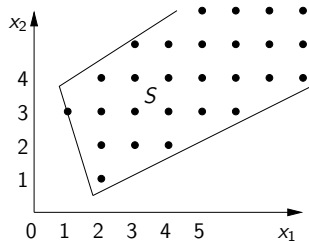
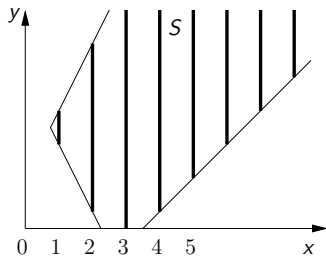
Mixed Integer Linear Programming Problems (MILP)

$$\begin{aligned} z_I &= \max c^\top x \\ Ax &\leq b \\ x &\geq 0 \\ x_i &\in \mathbb{Z}, \quad i \in I. \end{aligned}$$

where $A \in \mathbb{R}^{m \times n}$, $b \in \mathbb{R}^m$, $c \in \mathbb{R}^n$ and $I \subseteq \{1, \dots, n\}$.

- ▶ **integer variables:** $x_i, i \in I$.
- ▶ **continuous variables:** $x_i, i \notin I$.
- ▶ **Pure Integer Linear Programming Problem (ILP):** if all variables are integer.
- ▶ **Binary linear programming:** if all variables are binary (i.e. $x_i \in \{0, 1\}$).
- ▶ **Feasible region:** $X = \{x : Ax \leq b, x \geq 0, x_i \in \mathbb{Z} \text{ for } i \in I\}$.

Mixed Integer Linear Programming Problems



Mixed Integer Linear Programming Problems

$$\begin{aligned} z_I &= \max c^T x \\ Ax &\leq b \\ x &\geq 0 \\ x_i &\in \mathbb{Z}, \quad i \in I. \end{aligned}$$

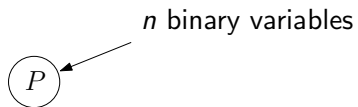
Linear relaxation of the problem

$$\begin{aligned} z_L &= \max c^T x \\ Ax &\leq b \\ x &\geq 0 \end{aligned} \tag{1}$$

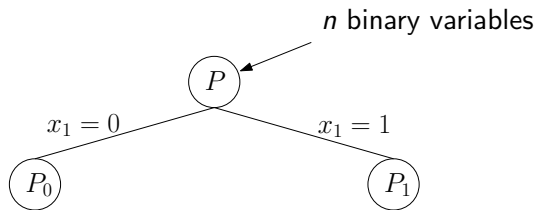
Note: $z_I \leq z_L$.

Observation: if the optimal solution of the linear relaxation x^L satisfies the integrality requirements, then x^L is optimal also for the (MILP) problem and $z_I = z_L$

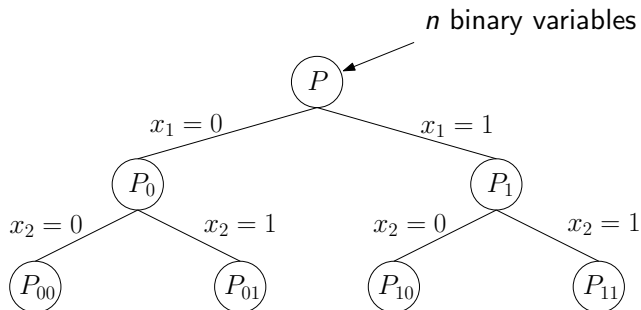
Enumeration by branching



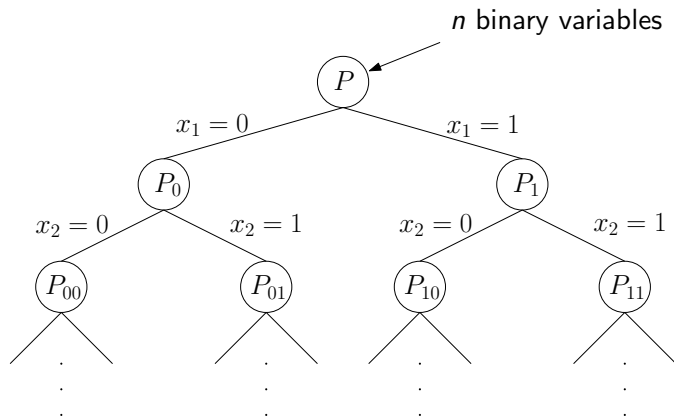
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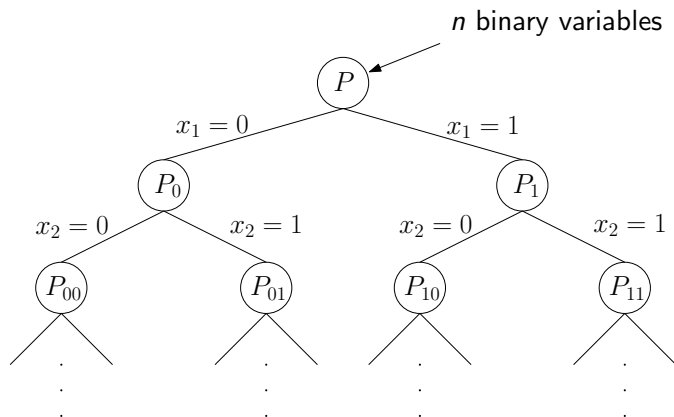
Enumeration by branching



Enumeration by branching



Enumeration by branching



2^n subproblems at the end

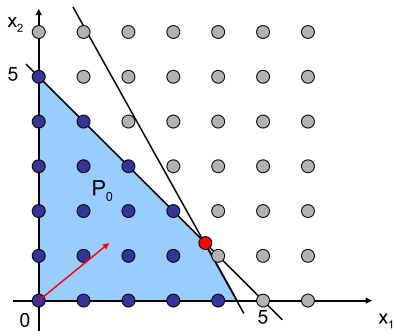
Drawbacks of complete enumeration

- ▶ Suppose that, in a IP with n binary variables, we enumerate all 2^n possible solutions.
- ▶ If we check one solution per microsecond:

# variables	Time
$n = 15$	$\sim .03$ sec
$n = 20$	~ 1 sec
$n = 30$	~ 20 min
$n = 40$	~ 12 days
$n = 50$	~ 35 years
$n = 60$	$\sim 37,500$ years
$n = 100$	$\sim 40,197$ trillion years

The branch-and-bound method

$$\begin{aligned} z_l^0 = \max \quad & 5x_1 + \frac{17}{4}x_2 \\ & x_1 + x_2 \leq 5 \\ & 10x_1 + 6x_2 \leq 45 \\ & x_1, x_2 \geq 0 \\ & x_1, x_2 \in \mathbb{Z} \end{aligned}$$



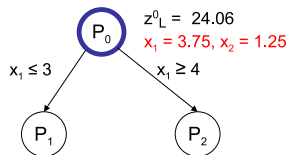
Optimal LP solution is $x_1 = 3.75$, $x_2 = 1.75$, with value $z_L^0 = 24.06$.

The branch-and-bound method

Branch on variable x_1 :

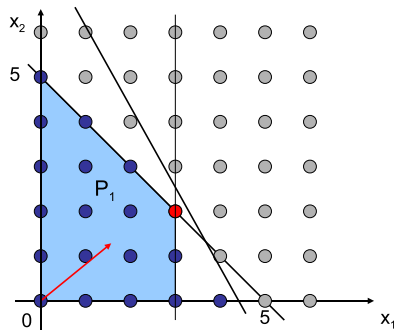
$$\begin{aligned} z_I^1 = \max \quad & 5x_1 + \frac{17}{4}x_2 \\ & x_1 + x_2 \leq 5 \\ & 10x_1 + 6x_2 \leq 45 \\ & x_1 \leq 3 \\ & x_1, x_2 \geq 0 \\ & x_1, x_2 \in \mathbb{Z} \end{aligned} \quad (P_1)$$

$$\begin{aligned} z_I^2 = \max \quad & 5x_1 + \frac{17}{4}x_2 \\ & x_1 + x_2 \leq 5 \\ & 10x_1 + 6x_2 \leq 45 \\ & x_1 \geq 4 \\ & x_1, x_2 \geq 0 \\ & x_1, x_2 \in \mathbb{Z} \end{aligned} \quad (P_2)$$



The branch-and-bound method

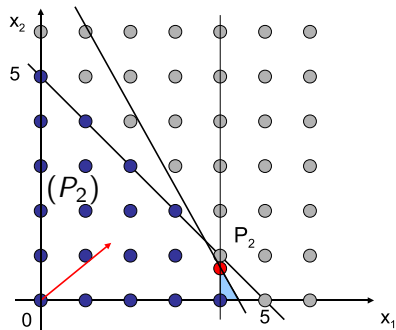
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Optimal LP solution $x_1 = 3$, $x_2 = 2$, with value $z_L^1 = 23.5$
Integer solution! $(3, 2)$ is the **incumbent solution**. $LB := 23.5$.
Prune node by optimality

The branch-and-bound method

$$\begin{aligned} z_I^1 = \max \quad & 5x_1 + \frac{17}{4}x_2 \\ & x_1 + x_2 \leq 5 \\ & 10x_1 + 6x_2 \leq 45 \\ & x_1 \geq 4 \\ & x_1, x_2 \geq 0 \\ & x_1, x_2 \in \mathbb{Z} \end{aligned}$$



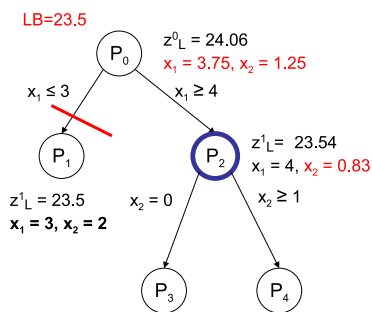
Optimal LP solution $x_1 = 4$, $x_2 = 0.83$, with value $z_L^2 = 23.54$.

The branch-and-bound method

Branch on variable x_2 :

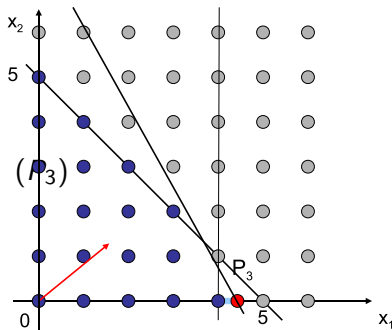
$$\begin{aligned} z_I^2 = \max \quad & 5x_1 + \frac{17}{4}x_2 \\ & x_1 + x_2 \leq 5 \\ & 10x_1 + 6x_2 \leq 45 \\ & x_1 \geq 4 \\ & x_2 \leq 0 \\ & x_1, x_2 \geq 0 \\ & x_1, x_2 \in \mathbb{Z} \end{aligned} \quad (P_3)$$

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The branch-and-bound method

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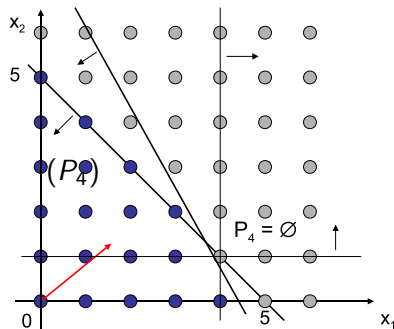


Optimal LP solution $x_1 = 4.5$, $x_2 = 0$, with value $z_L^3 = 22.5$.

$z_L^3 < LB(22.5 < 23.5) \implies$ **prune by bound.**

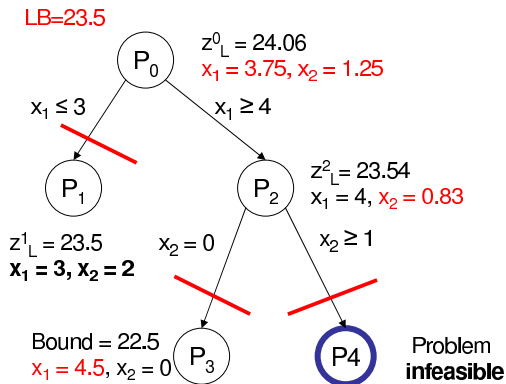
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LP relaxation infeasible: **prune by infeasibility.**

The branch-and-bound method



No more active nodes: the optimal solution is $(3, 2)$.

The branch-and-bound method (maximization)

Initialization: $\mathcal{T} := \{(P_0)\}$, $LB := -\infty$, x^* undefined;

1. If \mathcal{T} has no active node, return the incumbent solution x^* , STOP;
2. Else, choose an active node (P) in \mathcal{T} .
3. Solve the linear relaxation of (P) .
 - 3.1 **Pruning by infeasibility:** If LP relaxation is infeasible, prune (P) ;
Else, let \bar{x} be the optimum to the LP relaxation of (P) .
 - 3.2 **Pruning by bound:** If $c^\top \bar{x} \leq LB$, then prune (P) ;
Else
 - 3.3 **Pruning by optimality:** If \bar{x}_i is integer for all $i \in I$, set $x^* := \bar{x}$ and $LB := c^\top \bar{x}$ and prune node (P) .
 - 3.4 **Branching on variable:** If none of (a), (b), (c) occurs, choose an integer variable x_h such that $\bar{x}_h \notin \mathbb{Z}$ and branch on x_h :

$$(P') := (P) \cap \{x_h \leq \lfloor \bar{x}_h \rfloor\} \quad , \quad (P'') := (P) \cap \{x_h \geq \lceil \bar{x}_h \rceil\}.$$

Add (P') and (P'') to \mathcal{T} .

4. Return to 1.

Example: how to restart after branching

Suppose we solve the LP relaxation of a MIP in 5 variables with $x_1, x_2, x_3 \in \mathbb{Z}$, to obtain the optimal tableau

$$\begin{array}{rcccccccl} \max z & & & & & & & & \\ & & +x_2 & & +4x_4 & +x_5 & = & -9 \\ & & -\frac{1}{5}x_2 & +x_3 & -\frac{1}{5}x_4 & +\frac{1}{5}x_5 & = & \frac{1}{5} \\ x_1 & +\frac{1}{5}x_2 & & & -\frac{4}{5}x_4 & -\frac{1}{5}x_5 & = & \frac{9}{5} \\ & & & & & & & x_1, \dots, x_5 \geq 0 \end{array}$$

We branch on x_1 . The subproblems are $x_1 \leq 1$ and $x_1 \geq 2$.

How do we restart?

Example: how to restart after branching

max z

$$\begin{array}{cccccccl} z & & +x_2 & & +4x_4 & +x_5 & = & -9 \\ & & -\frac{1}{5}x_2 & +x_3 & -\frac{1}{5}x_4 & +\frac{1}{5}x_5 & = & \frac{1}{5} \\ x_1 & +\frac{1}{5}x_2 & & & -\frac{4}{5}x_4 & -\frac{1}{5}x_5 & = & \frac{9}{5} \\ & & & & & & & x_1, \dots, x_5 \geq 0 \end{array}$$

Subproblem: $x_1 \leq 1$

Add new slack variable $x_6 \geq 0$ and new inequality $x_1 + x_6 = 1$, expressed by non-basic variables as

$$-\frac{1}{5}x_2 + \frac{4}{5}x_4 + \frac{1}{5}x_5 + x_6 = -\frac{4}{5}.$$

Example: how to restart after branching

max z

$$\begin{array}{rcccccccl}
 z & & +x_2 & & +4x_4 & & +x_5 & & = & -9 \\
 & & -\frac{1}{5}x_2 & +x_3 & -\frac{1}{5}x_4 & & +\frac{1}{5}x_5 & & = & \frac{1}{5} \\
 x_1 & +\frac{1}{5}x_2 & & & -\frac{4}{5}x_4 & & -\frac{1}{5}x_5 & & = & \frac{9}{5} \\
 & -\frac{1}{5}x_2 & & & +\frac{4}{5}x_4 & & +\frac{1}{5}x_5 & +x_6 & = & -\frac{4}{5}
 \end{array}$$

$$x_1, \dots, x_5, x_6 \geq 0$$

New dual feasible basis: $\{1, 3, 6\}$.

Good formulations

MILP Formulations

$$\begin{aligned} z_I &= \max c^\top x \\ Ax &\leq b \\ x &\geq 0 \\ x_i &\in \mathbb{Z}, \quad i \in I. \end{aligned}$$

$$X = \{x \in \mathbb{R}^n : Ax \leq b, x \geq 0, x_i \in \mathbb{Z} \text{ for all } i \in I\}$$

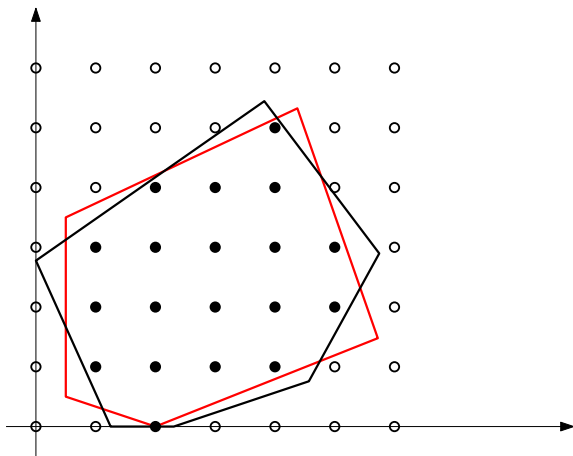
A **formulation** for the above (MILP) is a system

$$\begin{aligned} A'x &\leq b' \\ x &\geq 0 \end{aligned}$$

such that

$$X = \{x \in \mathbb{R}^n : A'x \leq b', x \geq 0, x_i \in \mathbb{Z} \text{ for all } i \in I\}.$$

MILP Formulations



MILP Formulations

Two formulations for X :

$$\begin{array}{l} Ax \leq b \\ x \geq 0 \end{array} \quad \text{and} \quad \begin{array}{l} A'x \leq b' \\ x \geq 0 \end{array}$$

Definition

The first formulation is **better** than the second if

$$\{x : Ax \leq b, x \geq 0\} \subseteq \{x : A'x \leq b', x \geq 0\}.$$

MILP Formulations

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$$\{x : Ax \leq b, x \geq 0\} \subseteq \{x : A'x \leq b', x \geq 0\}.$$

Note: if $Ax \leq b, x \geq 0$ is a better formulation than $A'x \leq b', x \geq 0$, then

$$z_I \leq z_L \leq z'_L.$$

Better formulations give tighter lower-bounds.

Knapsack problems

We are given a knapsack of capacity b , and n types of objects of positive weight a_1, \dots, a_n . Each object i has value c_i , $i = 1, \dots, n$.

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Knapsack problem: *Load the knapsack without exceeding capacity in order to maximize the total value carried.*

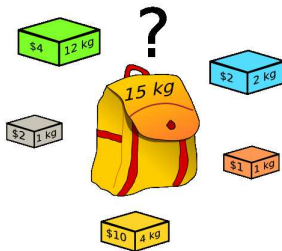
Binary knapsack problem: *Knapsack problem where we have exactly one object of each type.*

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Formulation for the knapsack problem

$$\begin{aligned} \max \quad & \sum_{i=1}^n c_i x_i \\ & \sum_{i=1}^n a_i x_i \leq b \\ & x \geq 0 \\ & x \in \mathbb{Z}^n \end{aligned}$$

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Formulation for the binary knapsack problem

$$\begin{aligned} \max \quad & \sum_{i=1}^n c_i x_i \\ & \sum_{i=1}^n a_i x_i \leq b \\ & 0 \leq x_i \leq 1 \\ & x \in \mathbb{Z}^n \end{aligned}$$

Knapsack problems: minimal cover formulation

Binary knapsack set

$$K := \{x \in \{0, 1\}^n : \sum_{i=1}^n a_i x_i \leq b\}$$

Covers: $C \subseteq \{1, \dots, n\}$ is a

- ▶ **cover** if $\sum_{i \in C} a_i > b$.
- ▶ **minimal cover** if it is a cover and $\sum_{i \in C \setminus \{j\}} a_i \leq b$ for every $j \in C$.

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Since objects in a cover exceed capacity, not all of them can be put in the knapsack. \Rightarrow Every point in K satisfy the

cover inequality: $\sum_{i \in C} x_i \leq |C| - 1$.

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Since objects in a cover exceed capacity, not all of them can be put in the knapsack. \Rightarrow Every point in K satisfy the

cover inequality: $\sum_{i \in C} x_i \leq |C| - 1$.

Observe that the **minimal cover inequalities** (i.e. the cover inequalities corresponding to minimal covers) imply all covers inequalities.

Knapsack cover inequalities

$$K^C := \{x \in \{0,1\}^n : \sum_{i \in C} x_i \leq |C|-1 \text{ for every minimal cover } C \text{ of } S\}.$$

Observation *The sets K and K^C coincide.*

Knapsack cover inequalities

$$K^C := \{x \in \{0,1\}^n : \sum_{i \in C} x_i \leq |C|-1 \text{ for every minimal cover } C \text{ of } S\}.$$

Observation *The sets K and K^C coincide.*

$$K := \{x \in \{0,1\}^4 : 6x_1 + 6x_2 + 6x_3 + 5x_4 \leq 16\}.$$

Minimal cover formulation:

$$K^C := \{x \in \{0,1\}^4 : \begin{array}{rrrr} x_1 & +x_2 & +x_3 & \leq 2 \\ x_1 & +x_2 & & +x_4 \leq 2 \\ x_1 & & +x_3 & +x_4 \leq 2 \\ & x_2 & +x_3 & +x_4 \leq 2 \end{array}\}.$$

In this instance, the second formulation is strictly better than the first.

Knapsack cover inequalities

$$K^C := \{x \in \{0,1\}^n : \sum_{i \in C} x_i \leq |C|-1 \text{ for every minimal cover } C \text{ of } S\}.$$

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Minimal cover formulation:

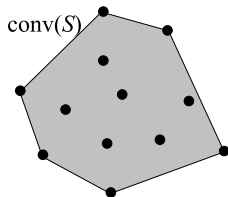
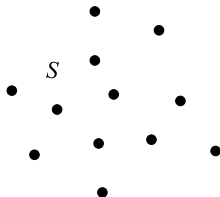
$$K^C := \{x \in \{0,1\}^4 : \begin{array}{rrrr} x_1 & +x_2 & +x_3 & \leq 2 \\ x_1 & +x_2 & & +x_4 \leq 2 \\ x_1 & & +x_3 & +x_4 \leq 2 \\ & x_2 & +x_3 & +x_4 \leq 2 \end{array}\}.$$

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$$(\frac{5}{6}, 1, 0, 1)$$

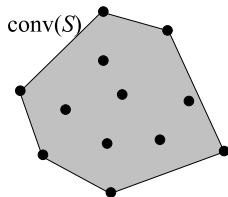
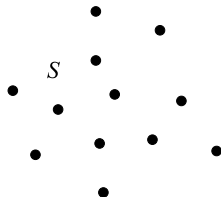
Ideal formulations

Given a set $X \subseteq \mathbb{R}^n$, the **convex hull** of X , denoted by $\text{conv}(X)$, is the (unique) minimal convex set containing X .



Ideal formulations

Given a set $X \subseteq \mathbb{R}^n$, the **convex hull** of X , denoted by $\text{conv}(X)$, is the (unique) minimal convex set containing X .



- ▶ $\text{conv}(X)$ is the intersection of all convex sets containing X .
- ▶ In particular, if K is a convex set and $X \subseteq K$, then $\text{conv}(X) \subseteq K$.

Ideal formulations

Given a set $X \subseteq \mathbb{R}^n$, the **convex hull** of X , denoted by $\text{conv}(X)$, is the (unique) minimal convex set containing X .



NOTE: If X is the set of feasible solutions of a mixed-integer program, and $Ax \leq b, x \geq 0$ is a formulation for X , then the polyhedron $P = \{x : Ax \leq b, x \geq 0\}$ satisfies

$$\text{conv}(X) \subseteq P.$$

Hence the best possible formulation of a MILP must contain $\text{conv}(X)$.

Ideal formulations

Theorem (Fundamental theorem of Integer Programming)

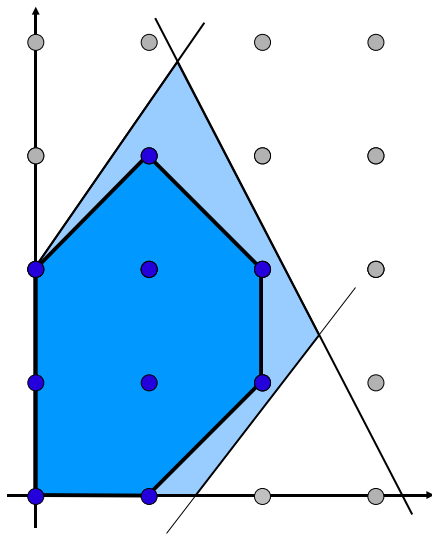
Given $A \in \mathbb{Q}^{m \times n}$, $b \in \mathbb{Q}^m$, let
 $X = \{x \in \mathbb{R}^n : Ax \leq b, x \geq 0, x_i \in \mathbb{Z} \text{ for } i \in I\}$. Then
 $\text{conv}(X)$ is a polyhedron.

\implies there exists $\tilde{A} \in \mathbb{Q}^{\tilde{m} \times n}$ and $\tilde{b} \in \mathbb{Q}^{\tilde{m}}$ such that

$$\text{conv}(X) = \{x \in \mathbb{R}^n : \tilde{A}x \leq \tilde{b}, x \geq 0\}.$$

$\tilde{A}x \leq \tilde{b}, x \geq 0$ is the ideal formulation for X .

Ideal formulations



Ideal formulations

Theorem

$X = \{x : Ax \leq b, x \geq 0, x_i \in \mathbb{Z} \text{ for } i \in I\}$. For any $c \in \mathbb{R}^n$, let

$$z_I = \max_{x \in X} c^\top x \quad \text{and} \quad \tilde{z} = \max_{x \in \text{conv}(X)} c^\top x \quad .$$

If all data is rational, then $z_I = \tilde{z}$, and *all vertices of $\text{conv}(X)$ are elements of X .*

Ideal formulations

Theorem

$X = \{x : Ax \leq b, x \geq 0, x_i \in \mathbb{Z} \text{ for } i \in I\}$. For any $c \in \mathbb{R}^n$, let

$$z_I = \max_{x \in X} c^\top x \quad \text{and} \quad \tilde{z} = \max_{x \in \text{conv}(X)} c^\top x \quad .$$

If all data is rational, then $z_I = \tilde{z}$, and *all vertices of $\text{conv}(X)$ are elements of X .*

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 - ▶ How do you find the ideal formulation?
 - ▶ The ideal formulation can be extremely complicated, and can have exponentially many constraints.