

Mixed Integer Non Linear Optimization: Methods and Applications

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Linear Programming

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Linear Programming problems

$$\begin{aligned} \min_x \quad & f(x) \\ & g_i(x) \leq 0 \quad \forall i = 1, \dots, m \\ \underline{x} \leq x \leq \bar{x} \\ & x_j \in \mathbb{Z} \quad \forall j \in Z \end{aligned}$$

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Moreover, $\underline{x} \in [-\infty, +\infty)$ and $\bar{x} \in (-\infty, +\infty]$.

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Geometrical interpretation of LPs

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Geometrical interpretation of LPs

How to draw constraints and objective function

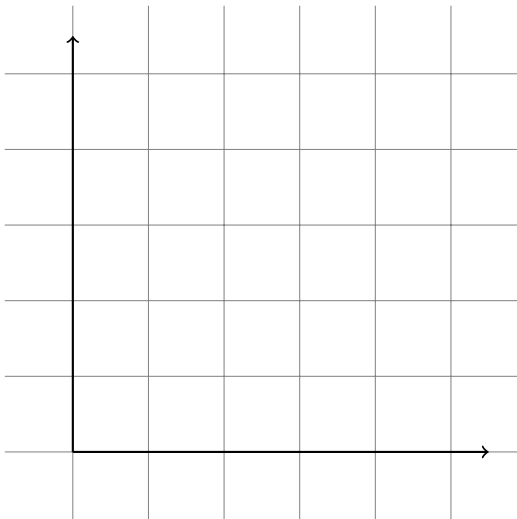
Example 1

$$\max x_1 + x_2$$

$$3x_1 + 2x_2 \leq 13$$

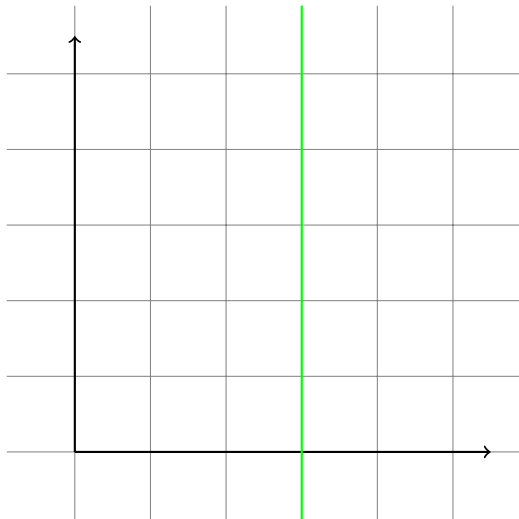
$$0 \leq x_1 \leq 3$$

$$0 \leq x_2 \leq 5.$$



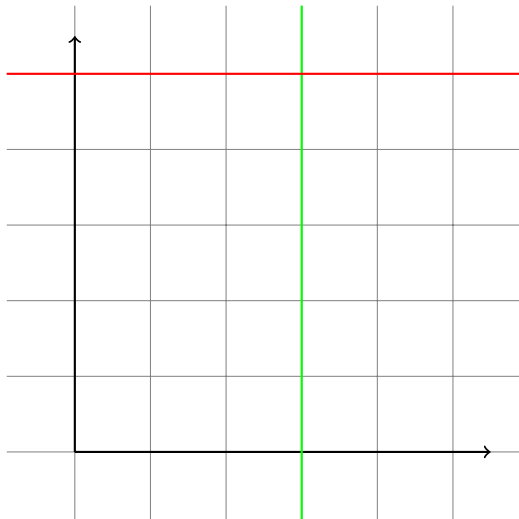
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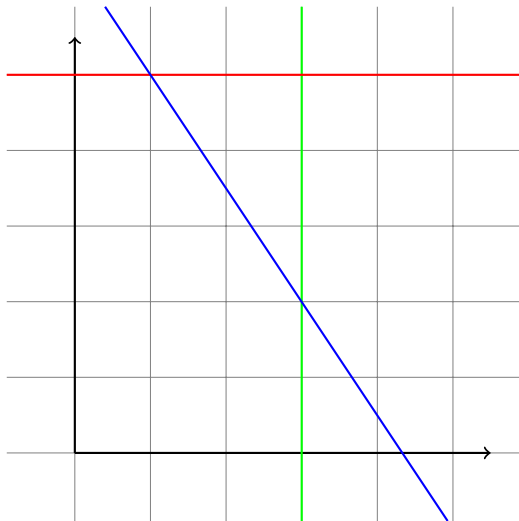
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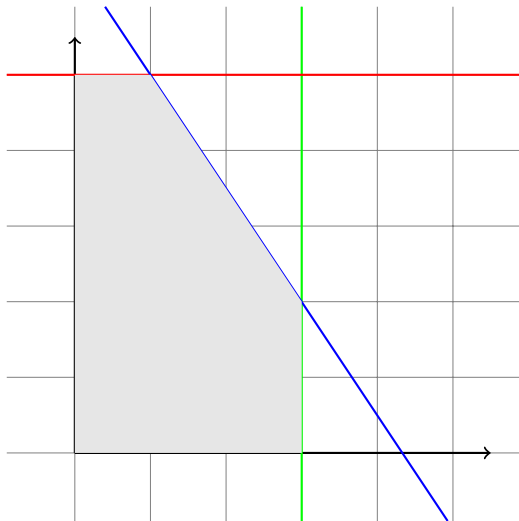
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A few definitions



Consider some $i \in \{1, \dots, m\}$

► **Hyperplane** : $\{x \in \mathbb{R}^n \mid A_i^\top x = b_i\}$

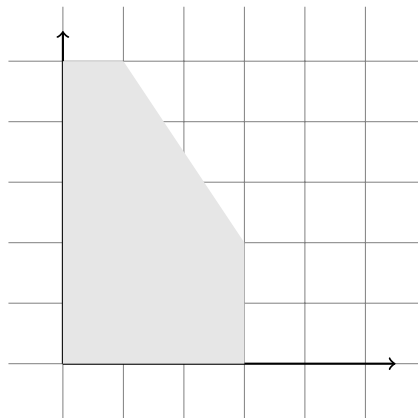
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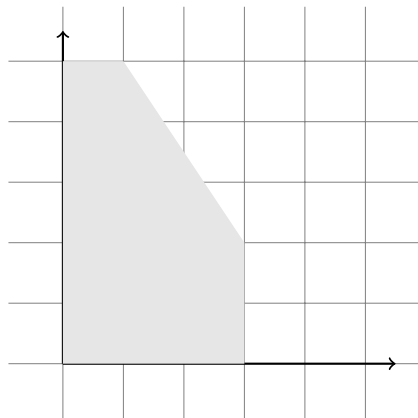
- **Hyperplane** : $\{x \in \mathbb{R}^n \mid A_i^\top x = b_i\}$
- **Half-space** : $\{x \in \mathbb{R}^n \mid A_i^\top x \leq b_i\}$ or $\{x \in \mathbb{R}^n \mid A_i^\top x \geq b_i\}$

A few definitions



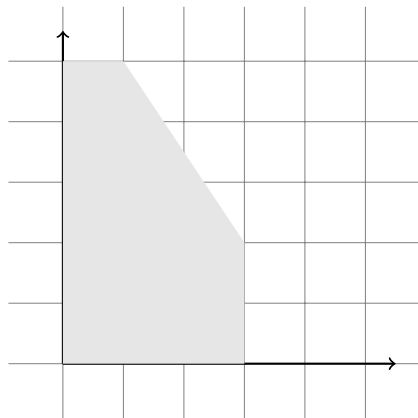
► Polyhedron : $\{x \in \mathbb{R}^n \mid Ax \leq b\}$

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Remark

The feasible region of a LP problem is a polyhedron (by definition).

Some properties and theorems

Definition

Given points $v^1, v^2, \dots, v^p \in \mathbb{R}^n$, their **convex combination** is $z = \sum_{i=1}^p \alpha_i v^i$ s.t. $\sum_{i=1}^p \alpha_i = 1$ and $\alpha_i \geq 0$ for all $i = 1, \dots, p$.

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Theorem

Every polyhedron $P \subseteq \mathbb{R}^d$ can be written as

$$P = \text{conv}\{v^1, \dots, v^k\} + \text{cone}\{r^1, \dots, r^\ell\}$$

with points $v^1, \dots, v^k \in \mathbb{R}^d$ and rays $r^1, \dots, r^\ell \in \mathbb{R}^d$

where

$$\text{cone}\{r^1, \dots, r^\ell\} = \{x \in \mathbb{R}^d \mid x = \mu_1 r^1 + \dots + \mu_\ell r^\ell, \mu_1, \dots, \mu_\ell \geq 0\}.$$

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Thus, a polytope can be **characterized/described** by a finite number of half-spaces (H-description) or its vertices (V-description).