

Hands-on Tutorial on Optimization

F. Eberle, R. Hoeksma, and N. Megow

September 24, 2019

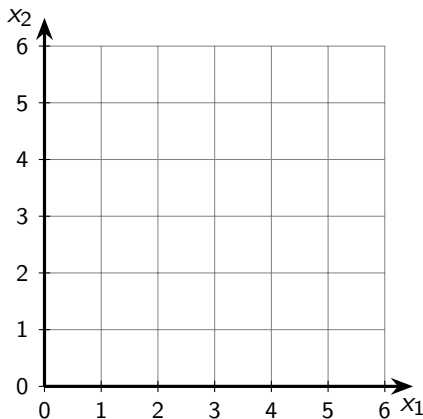
# Geometry of Linear Programs

# The Real, $n$ -dimensional Space

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$$\mathbb{R}^n := \{x = (x_1, \dots, x_n) : x_1, \dots, x_n \in \mathbb{R}\}$$

Elements  $x \in \mathbb{R}^n$  can be seen as



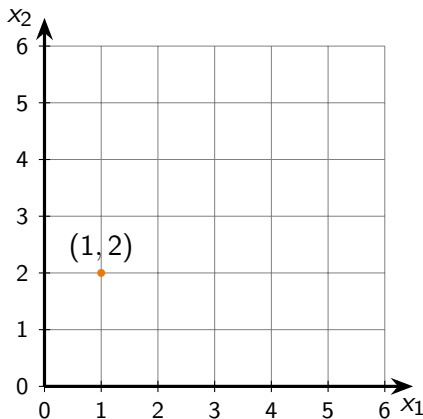
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► Points



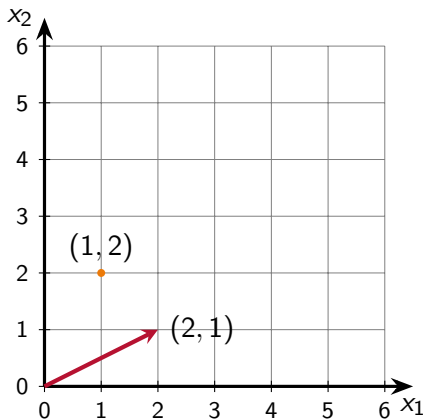
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Elements  $x \in \mathbb{R}^n$  can be seen as

- ▶ Points
- ▶ Vectors



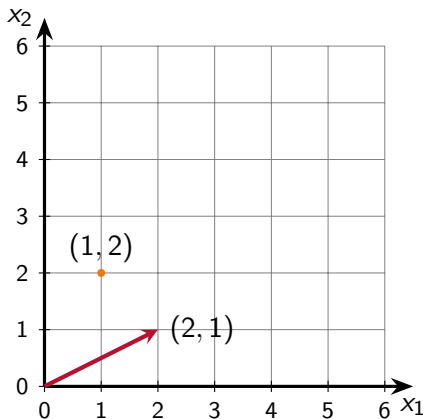
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Elements  $x \in \mathbb{R}^n$  can be seen as

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- ▶ Vectors

A  $n$ -tuple may represent the net profit of  $n$  different goods, or their inventory level, or the cost of production, etc.

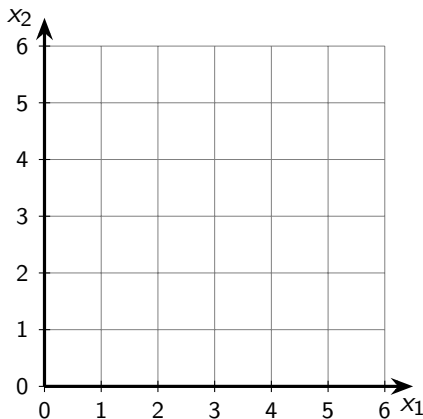


# Linear Equations

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Let  $a \in \mathbb{R}^n$  be a profit vector.

If you want the profit to be exactly  $b$ , how much should you produce?



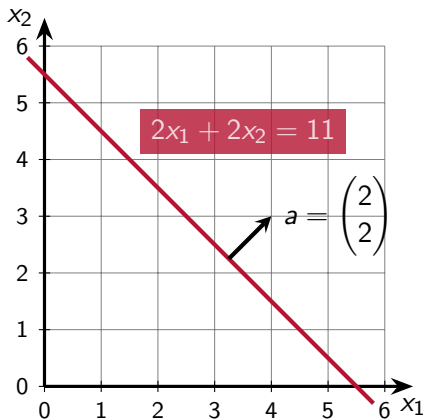
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This is a line in  $\mathbb{R}^n$ .



# Linear Equations

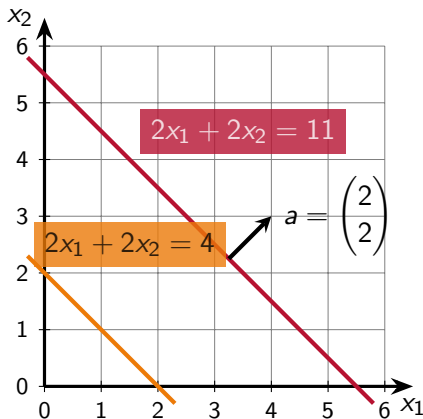
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This is a line in  $\mathbb{R}^n$ .

Changing the right hand side  $b$  corresponds to “moving” the line along the vector  $a$ .



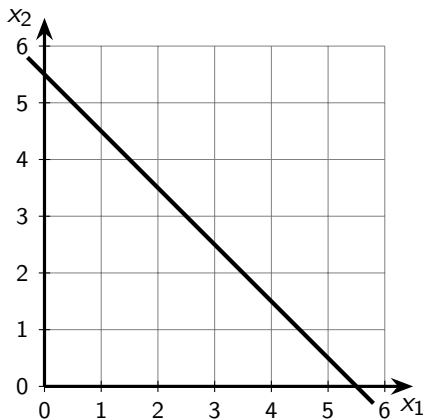


# Linear Inequalities

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If you want to earn **at least** (at **most**)  $b$ , how much should you produce?



# Linear Inequalities

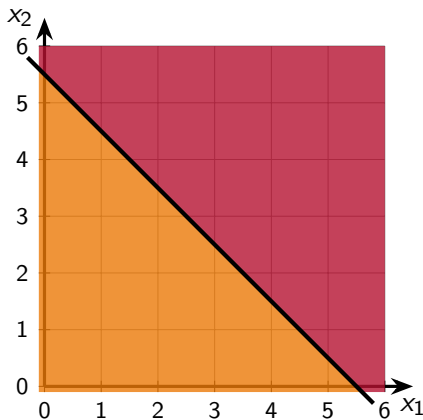
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The answer is given as the set  $\{x \in \mathbb{R}^n : a_1x_1 + \dots + a_nx_n \geq (\leq)b\}$ .

This is a halfspace in  $\mathbb{R}^n$ .

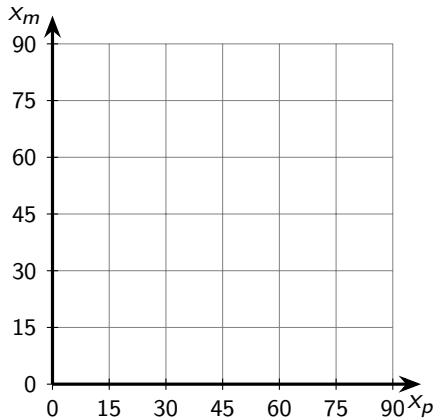


# Graphical Solution of LPs

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Consider the chips factory problem.

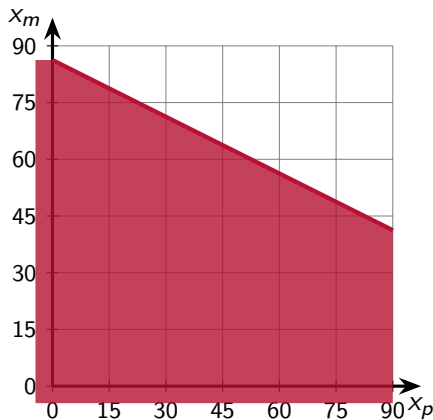
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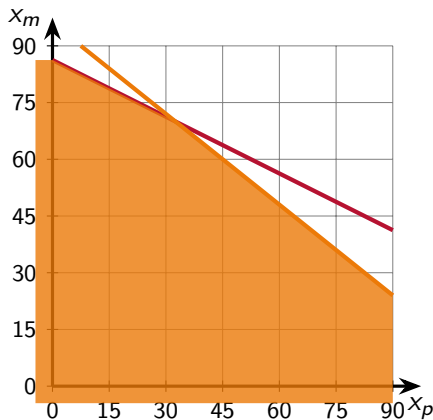
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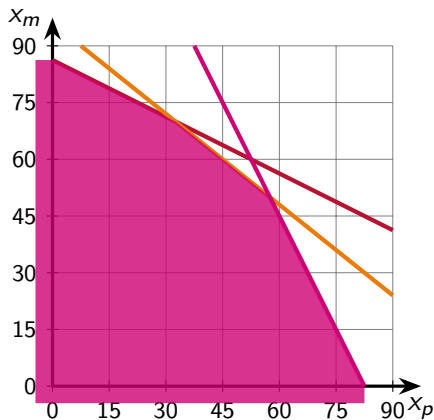
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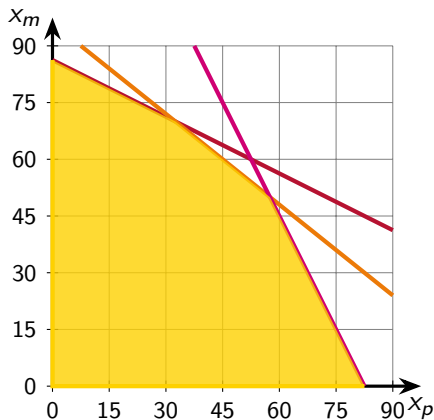
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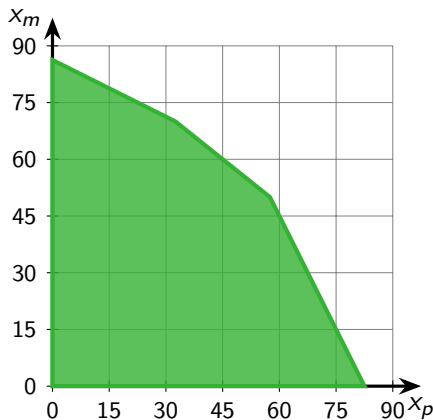
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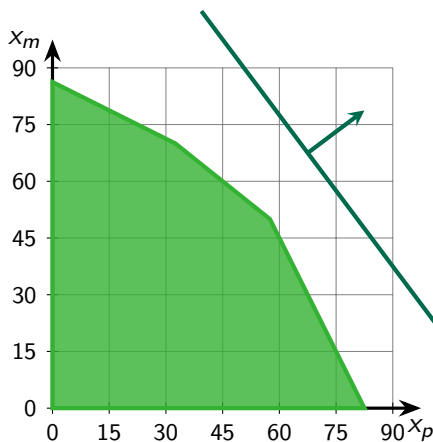




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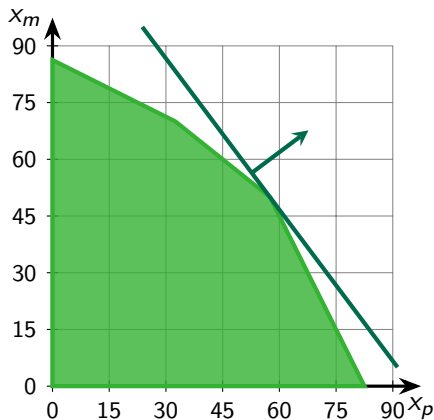
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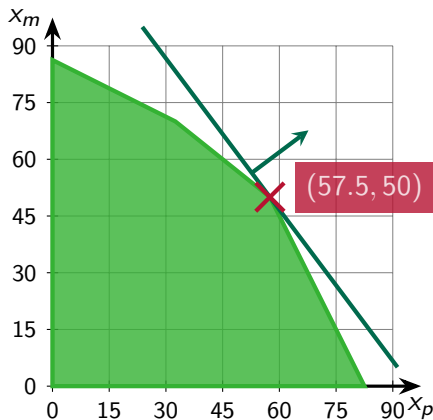
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Optimal solution:

$$(x_p, x_m) = (57.5, 50), \text{ s.t.} \\ 2 \cdot 57.5 + \frac{3}{2} \cdot 50 = 190.$$



## Graphical Solution of LPs

---

Consider the crude oil processing problem.

Processes:

Process	cost per barrel	output per 10 barrels
1	3€	2 barrels heavy oil 2 barrels medium heavy oil 1 barrel light oil
2	5€	1 barrel heavy oil 2 barrels medium heavy oil 4 barrels light oil

Demands: heavy oil: 3 barrels  
medium heavy oil: 5 barrels  
light oil: 4 barrels

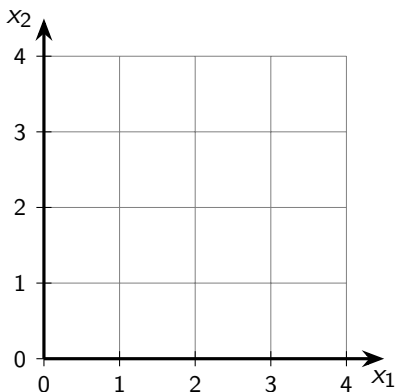
Construct the polyhedral representation and solve the problem then graphically by hand.

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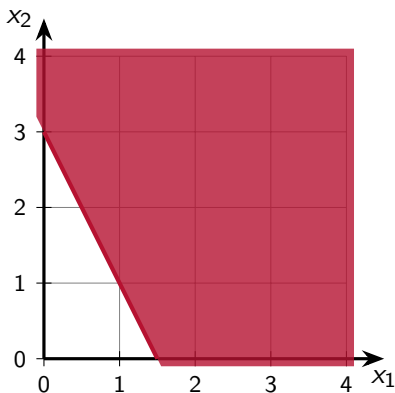
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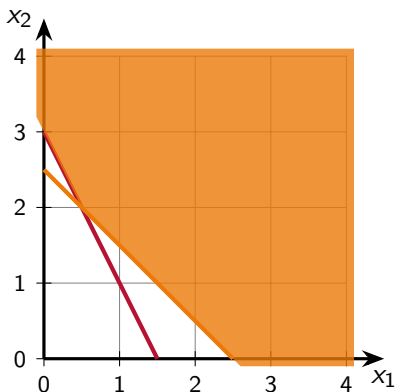
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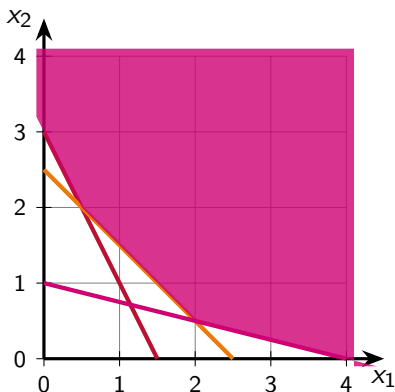
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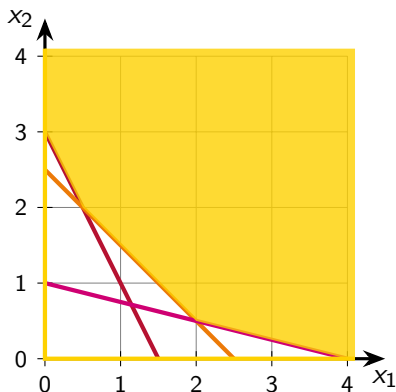




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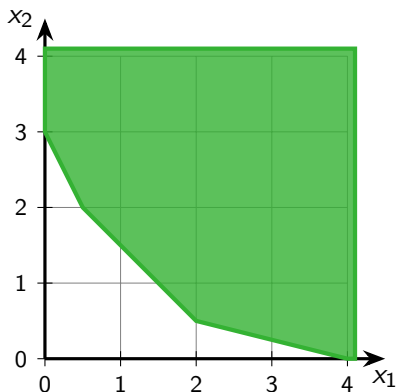
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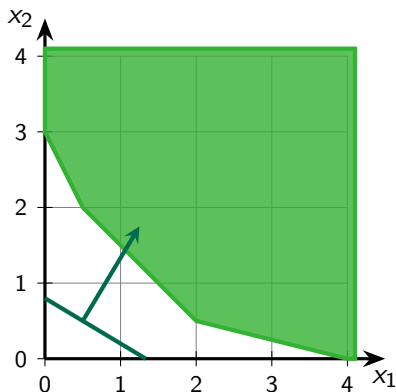
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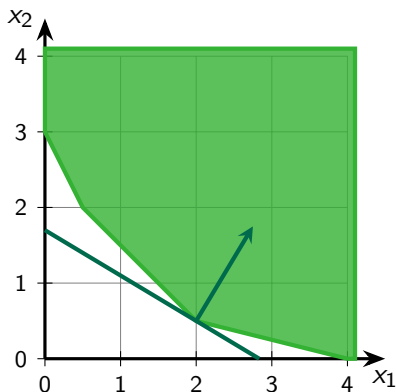
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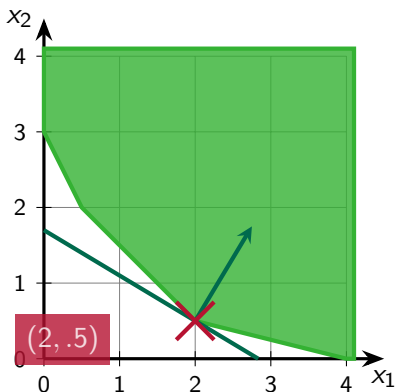


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Optimal solution:  $(x_1, x_2) = (2, .5)$ ,  
s.t.  $30 \cdot 2 + 50 \cdot .5 = 85$ .



# Extreme points

---

## Definition

Let  $M \neq \emptyset$  convex. A point  $x \in M$  is an **extreme point** of  $M$  if we cannot find two points  $y, z \in M \setminus \{x\}$  and a scalar  $\lambda \in (0, 1)$  such that

$$x = \lambda y + (1 - \lambda)z$$

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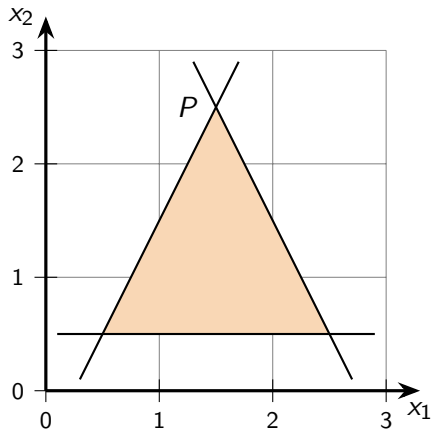
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- ▶ Convex sets with finitely many extreme points are **polytopes** (bounded) or **polyhedra** (unbounded).
- ▶ Examples of convex sets with infinitely many extreme points: circle or ball

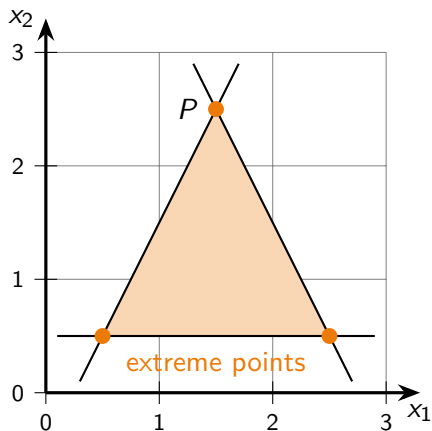
## Example

$$P = \left\{ x \in \mathbb{R}^2 : \begin{array}{l} x_2 \geq \frac{1}{2} \\ 2x_1 + x_2 \leq \frac{11}{2} \\ -2x_1 + x_2 \leq -\frac{1}{2} \end{array} \right\}$$



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# Matrices

---

A  $m \times n$  dimensional matrix  $A$  is an array of  $n \cdot m$  real numbers  $a_{i,j}$ :

$$A = \begin{pmatrix} a_{1,1} & a_{1,2} & \cdots & a_{1,n} \\ a_{2,1} & a_{2,2} & \cdots & a_{2,n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m,1} & a_{m,2} & \cdots & a_{m,n} \end{pmatrix} \in \mathbb{R}^{m \times n}$$

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The transposed of  $A$  is defined as follow:

$$A^T := \begin{pmatrix} a_{1,1} & a_{2,1} & \cdots & a_{m,1} \\ a_{1,2} & a_{2,2} & \cdots & a_{m,2} \\ \vdots & \vdots & \ddots & \vdots \\ a_{1,n} & a_{2,n} & \cdots & a_{m,n} \end{pmatrix} \in \mathbb{R}^{n \times m}$$

# Special Matrices

---

Let  $\mathbf{0} \in \mathbb{R}^{m \times n}$  denote the matrix that only contains 0.

By  $\mathbf{1} \in \mathbb{R}^{n \times n}$  we denote the matrix that only consists of 0s except the diagonal:

$$\mathbf{1} := \begin{pmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 1 \end{pmatrix} \in \mathbb{R}^{n \times n}$$

# Multiplication

---

Let  $x, c \in \mathbb{R}^n$ ,  $A \in \mathbb{R}^{m \times n}$ , and  $b \in \mathbb{R}^m$ . Then,  $c^T x := \sum_{j=1}^n c_j \cdot x_j$ , and

$$Ax = \begin{pmatrix} a_{1,1} & a_{1,2} & \cdots & a_{1,n} \\ a_{2,1} & a_{2,2} & \cdots & a_{2,n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m,1} & a_{m,2} & \cdots & a_{m,n} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} := \begin{pmatrix} \sum_{j=1}^n a_{1,j} \cdot x_j \\ \sum_{j=1}^n a_{2,j} \cdot x_j \\ \vdots \\ \sum_{j=1}^n a_{m,j} \cdot x_j \end{pmatrix} \in \mathbb{R}^m$$

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Thus,

$$\max c^T x$$

$$Ax \leq b$$



$$\max \sum_{j=1}^n c_j \cdot x_j$$

$$\sum_{j=1}^n a_{1,j} \cdot x_j \leq b_1$$

$$\sum_{j=1}^n a_{2,j} \cdot x_j \leq b_2$$

$$\vdots \quad \quad \quad \vdots$$

$$\sum_{j=1}^n a_{m,j} \cdot x_j \leq b_3$$



## Example

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$$P = \left\{ x \in \mathbb{R}^2 : \begin{array}{rcl} & x_2 & \geq \frac{1}{2} \\ 2x_1 + & x_2 & \leq \frac{11}{2} \\ -2x_1 + & x_2 & \leq -\frac{1}{2} \end{array} \right\}$$
$$= \left\{ x \in \mathbb{R}^2 : \begin{pmatrix} 0 & -1 \\ 2 & 1 \\ -2 & 1 \end{pmatrix} x \leq \begin{pmatrix} -\frac{1}{2} \\ \frac{11}{2} \\ -\frac{1}{2} \end{pmatrix} \right\}$$

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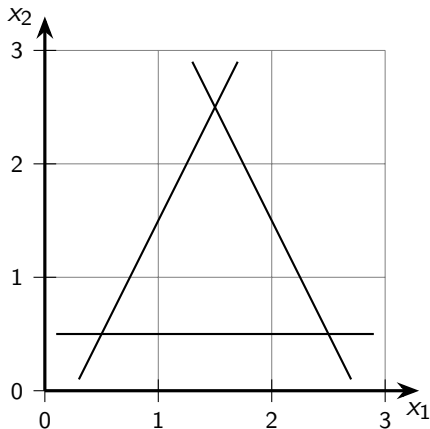
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- ▶  $P \subseteq \mathbb{R}^n$  is a polyhedron if  $P$  is the intersection of finitely many closed halfspaces.

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- ▶ The set  $\{x \in \mathbb{R}^n \mid Ax \leq (\geq) b\}$  is called **polyhedron**.
- ▶ The set  $\{x \in \mathbb{R}^n \mid Ax = b, x \geq 0\}$  is a **polyhedron in standard form**.
- ▶  $P \subseteq \mathbb{R}^n$  is a polyhedron if  $P$  is the intersection of finitely many closed halfspaces.
- ▶ A bounded polyhedron is called **polytop**.

## Example

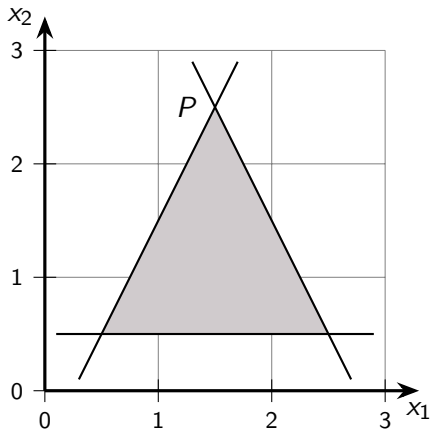
$$P = \left\{ x \in \mathbb{R}^2 : \begin{array}{l} x_2 \geq \frac{1}{2} \\ 2x_1 + x_2 \leq \frac{11}{2} \\ -2x_1 + x_2 \leq -\frac{1}{2} \end{array} \right\}$$



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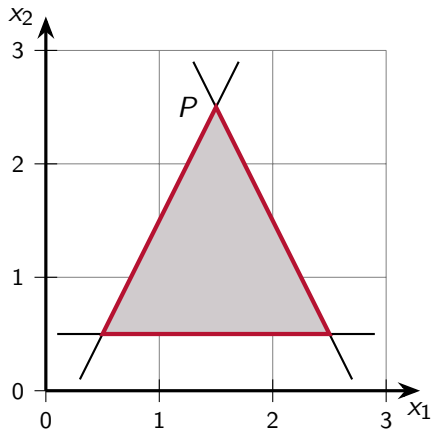


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Edges of  $P$





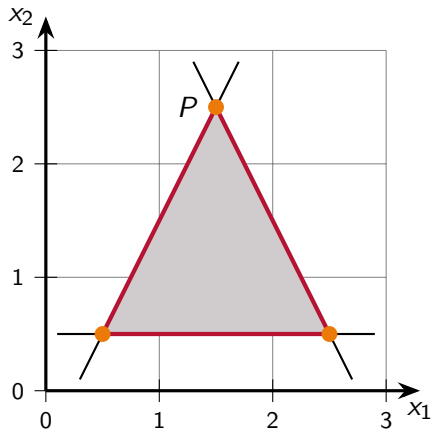
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Facet  $P$

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Vertices of  $P$



# Optimal Solutions

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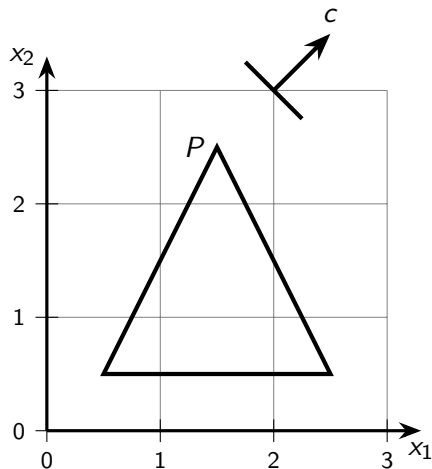
Let  $P = P(A, b) = \{x \in \mathbb{R}^n \mid Ax \leq b\} \neq \emptyset$  a polyhedron.

## Theorem

The optimum of  $\min\{c^T x \mid x \in P\}$  is attained at a vertex of  $P$ .

# Example

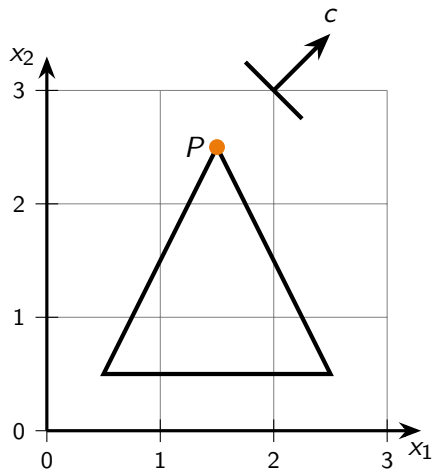
$$\begin{array}{llll} \min & -x_1 & - & x_2 \\ \text{s.t.} & & & x_2 \geq \frac{1}{2} \\ & 2x_1 & + & x_2 \leq \frac{11}{2} \\ & -2x_1 & + & x_2 \leq -\frac{1}{2} \end{array}$$



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Optimal solution



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**Example:** Unit cube  $P = \{x \in \mathbb{R}^n \mid 0 \leq x_i \leq 1, i = 1, \dots, n\}$

Number of inequalities:  $m = 2n$

Number of vertices:  $2^n$

Works well in practice.





**Break**