



## Definition

An optimization problem  $(\mathcal{X}, f)$  is called *linear* (a *linear optimization problem*) if  $\mathcal{X} \subset \mathbf{R}^n$  is a polyhedron and  $f : \mathbf{R}^n \rightarrow \bar{\mathbf{R}}$  is a linear function.

## Problem data

Recall that  $f$  is linear means there exists  $c \in \mathbf{R}^n$  such that

$$f(x) = c^\top x \quad \text{for all } x \in \mathbf{R}^n$$

Also,  $\mathcal{X}$  polyhedral means there exists  $A \in \mathbf{R}^{m \times n}$  and  $b \in \mathbf{R}^d$  such that

$$\mathcal{X} = \{x \in \mathbf{R}^n \mid Ax \leq b\}$$

For this reason, the *problem data*  $(A, b, c)$  is sufficient to specify a linear optimization problem. Recall that  $Ax \leq b$  means element-wise inequality (i.e., that the inequality holds in each component).

## Task

Given data  $A \in \mathbf{R}^{m \times n}$ ,  $b \in \mathbf{R}^n$ ,  $c \in \mathbf{R}^n$ , we want to find  $x \in \mathbf{R}^d$  to

$$\begin{aligned} &\text{minimize} && c^\top x \\ &\text{subject to} && Ax \leq b \end{aligned}$$

We either want  $x^* \in \mathbf{R}^n$  so that  $Ax^* \leq b$  and  $c^\top x^* \leq c^\top x$  for all  $x \in \mathbf{R}^n$ , or we want to know that  $\{x \mid Ax \leq b\} = \emptyset$ , or we want to know that for all  $\alpha \in \mathbf{R}$ , there is an  $x \in \mathbf{R}^n$  satisfying  $Ax \leq b$  and  $c^\top x \leq \alpha$ . This problem is regularly called *linear programming* (a *linear program*). Many authors define this problem with the goal as maximization: of course, minimizing  $c^\top x$  is equivalent to (has the same set of optimal solutions) maximizing  $-c^\top x$ .

## Notation

In the context of linear optimization,  $c^\top x$  is often abbreviated  $cx$ . A linear program is sometimes abbreviated  $\min\{cx \mid Ax \leq b\}$  (here the

matrices and vectors are assumed to be conforming). As usual,  $x \in \mathbf{R}^n$  is called *feasible* (a *feasible solution*) if  $Ax \leq b$ .  $x^* \in \mathbf{R}^n$  is called *optimal* (an *optimal solution*, *optimum solution*) if  $c^\top x^* \leq c^\top x$  for all  $x \in \mathbf{R}^n$ . We sometimes denote the rows of  $a$  by  $\bar{a}_i^\top \in \mathbf{R}^n$  for  $i = 1, \dots, m$ , i.e.,

$$A = \begin{bmatrix} - & \bar{a}_1^\top & - \\ & \vdots & \\ - & \bar{a}_m^\top & - \end{bmatrix} \in \mathbf{R}^{m \times n}$$

and refer to the inequality  $\bar{a}_i^\top x \leq b_i$  as an *inequality constraint* for  $i = 1, \dots, m$ . The set or the expression  $Ax \leq b$  are both sometimes called the *inequality constraints* of the problem.

The problem  $(A, b, c)$  is *infeasible* if the polyhedron

$$P = \{x \in \mathbf{R}^n \mid Ax \leq b\}$$

is empty. In symbols, if  $P = \emptyset$ . The problem is *unbounded* if for all  $\alpha \in \mathbf{R}$ , there exists  $x \in P$  with  $c^\top x < \alpha$ .

Here's an infeasible instance. Define

$$A = \begin{bmatrix} 1 \\ -1 \end{bmatrix} \quad b = \begin{bmatrix} -1 \\ -1 \end{bmatrix} \quad c = \begin{bmatrix} 1 \end{bmatrix}$$

We want  $x \in \mathbf{R}^1$  so that  $x \leq -1$  and  $x \geq 1$ . There is no such  $x$ .

Here's an unbounded instance: drop the second inequality constraint. Then we are interested in finding  $x_1$  to minimize  $x_1$  subject to  $x_1 \leq -1$ . Given  $\alpha \in \mathbf{R}$ , if  $\alpha > 0$  pick  $x_1 = -1$ , else pick  $2\alpha$ . This problem is unbounded.

Here's a simple example with an optimal solution. Modify  $b = \begin{bmatrix} 1 & 0 \end{bmatrix}^\top$ . Now we want to find  $x_1$  so that

$$x_1 \leq 1 \quad \text{and} \quad x_1 \geq 0$$

and we minimize  $x_1$ . Clearly  $x_1 = 0$  is an optimal solution. Indeed, it is the unique optimal solution in this case.

