# Modeling linear programming problems

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## **Outline**

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Motivation

# What is optimization?

(Merriam-Webster Dictionary) An act, process, or methodology of making something (such as a design, system, or decision) as fully perfect, functional, or effective as possible.

Motivation

## (Maximum Area Problem)

You have 80 meters of wire and want to enclose a rectangle as large as possible (in area). How should you do it?

Note: This is a non-linear optimization problem!

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## **Common Framework**

Components of an optimization problem

- Decisions
- Constraints
- Objective

Optimization seeks to choose some decisions to optimize (maximize or minimize) an objective subject to certain constraints.

## **Common Framework**

Given  $f, g_i, h_i : \mathbb{R}^n \mapsto \mathbb{R}$ 

$$Z = \text{minimize}/\text{maximize}$$
  $f(\mathbf{x})$  (1a)

subject to 
$$g_i(\mathbf{x}) \le 0, \forall i = 1, 2, ..., p$$
 (1b)

$$g_j(\mathbf{x}) \ge 0, \forall j = 1, 2, ..., q$$
 (1c)

$$h_k(\mathbf{x}) = 0, \forall k = 1, 2, ..., r$$
 (1d)

- ▶ Decisions:  $\mathbf{x}$ , Objective:  $f(\mathbf{x})$ , and Constraints: (1b)-(1d)
- $\blacktriangleright$  (1b), (1c), and (1d): set of " $\leq$ ", " $\geq$ ", and equality constraints
- $ightharpoonup \mathcal{X} = \{\mathbf{x} \in \mathbb{R}^n : (1\mathsf{b}) (1\mathsf{d})\}$  define the feasible region.
- Any  $\hat{\mathbf{x}}$  satisfying all the constraints is a feasible solution.
- ▶ Any  $\mathbf{x}^* \in \mathcal{X}$  satisfying  $f(\mathbf{x}^*) \leq f(\mathbf{x}), \forall \mathbf{x} \in \mathcal{X}$  is an optimal solution.
- $lackbox{ } f(\mathbf{x}^*)$  is known as optimal objective value.

# A few classes of optimization problems

- Linear optimization:  $f, g_i, h_i$  are all affine functions of continuous variables x.
- Non-linear optimization: At least one of  $f, g_i, h_i$  is non-linear function of continuous variables x.
  - Convex optimization: All functions are convex and feasible region is a convex set
- lacktriangle (Mixed) Integer optimization: Some of the variables x are restricted to be integers.
- lacktriangleright (Mixed) Integer Non-linear optimization: Some of the variables x are restricted to be integers and at least one of  $f,g_i,h_i$  is non-linear.

Difficulty of solving above classes rises significantly as we go from above to below.

## A few definitions

Definition (Maximum) Let  $S \subseteq \mathbb{R}$ . We say that x is a maximum of S iff  $x \in S$  and  $x \geq y, \forall y \in S$ .

Definition (Minimum) Let  $S \subseteq \mathbb{R}$ . We say that x is a minimum of S iff  $x \in S$  and  $x \leq y, \forall y \in S$ .

Definition (Bounds) Let  $S \subseteq \mathbb{R}$ . We say that u is an upper bound of S iff  $u \ge x, \forall x \in S$ . Similarly, l is a lower bound of S iff  $l \le x, \forall x \in S$ .

Definition (Supremum) Let  $S\subseteq\mathbb{R}$ . We define the supremum of S denoted by  $\sup(S)$  to be the smallest upper bound of S. If no such upper bound exists, then we set  $\sup(S)=+\infty$ .

Definition (Infimum) Let  $S\subseteq\mathbb{R}$ . We define the infimum of S denoted by  $\inf(S)$  to be the largest lower bound of S. If no such lower bound exists, then we set  $\inf(S)=-\infty$ 

Definition If  $x \in S$  such that  $x = \sup(S)$ , we say that supremum of S is achieved (which means that there is a maximum to the problem). Similar definition for whether infimum is achieved.

### **General Formulation of LP**

$$Z = \underset{\mathbf{x}}{\text{minimize}} / \underset{\mathbf{x}}{\text{maximize}} \qquad \mathbf{c}^T \mathbf{x} \qquad \qquad (2a)$$
 subject to 
$$\mathbf{a}_i^T \mathbf{x} \leq b_i, \forall i \in C_1 \qquad (2b)$$
 
$$\mathbf{a}_j^T \mathbf{x} \geq b_j, \forall j \in C_2 \qquad (2c)$$
 
$$\mathbf{a}_k^T \mathbf{x} = b_k, \forall k \in C_3 \qquad (2d)$$
 
$$x_u \geq 0, \forall u \in N_1 \qquad (2e)$$
 
$$x_v \leq 0, \forall v \in N_2 \qquad (2f)$$
 
$$x_w \text{ free }, \forall w \in N_3 \qquad (2g)$$

where,  $C_1, C_2, C_3 \subseteq \{1,...,m\}$ ,  $N_1, N_2, N_3 \subseteq \{1,...,n\}$ 

#### More definitions

Definition (Hyperplane)  $\{\mathbf{x} \in \mathbb{R}^n : \mathbf{a}^T \mathbf{x} = b\}$ 

Definition (Halfspace)  $\{\mathbf{x} \in \mathbb{R}^n : \mathbf{a}^T \mathbf{x} \geq b\}$ 

Definition (Polyhedron) A set  $P \subseteq \mathbb{R}^n$  is called a polyhedron if P is the intersection of a finite number of halfspaces.  $P = \{\mathbf{x} \in \mathbb{R}^n : A\mathbf{x} \leq \mathbf{b}\}$ 

Definition (Polytope) A bounded polyhedron is called a polytope.

Question Is  $\{\mathbf{x} \in \mathbb{R}^n : A\mathbf{x} = \mathbf{b}, \mathbf{x} \ge 0\}$  a polyhedron?

Definition (Convex Sets) A set  $S \subseteq \mathbb{R}^n$  is a convex set if for any  $\mathbf{x}, \mathbf{y} \in S$ , and  $\lambda \in [0,1]$ , we have  $\lambda \mathbf{x} + (1-\lambda)\mathbf{y} \in S$ . Question Is polyhedron  $P = \{\mathbf{x} \in \mathbb{R}^n : A\mathbf{x} \leq \mathbf{b}\}$  a convex set?

Definition (Convex combination)  $\mathbf{x} \in \mathbb{R}^n$  is said to be convex combination of  $\mathbf{x}^1,...,\mathbf{x}^p \in \mathbb{R}^n$  if for  $\lambda_1,...,\lambda_p \geq 0$  s.t.  $\sum_i^n \lambda_i = 1$ ,  $\mathbf{x}$  can be expressed as  $\mathbf{x} = \sum_i^n \lambda_i \mathbf{x}^i$ .

Definition (Extreme point) Let P be a polyhedron. Then,  $\mathbf{x} \in P$  is an extreme point of P if we cannot express  $\mathbf{x}$  as a convex combination of other points in P.

#### **Theorem**

Let P be a non-empty polyhedron. Consider  $\mathsf{LP} \max \{ \mathbf{c}^T \mathbf{x} \ s.t. \ \mathbf{x} \in P \}$ . Suppose the  $\mathsf{LP}$  has at least one optimal solution and P has at least one extreme point. Then, above  $\mathsf{LP}$  has at least one extreme point of P that is an optimal solution.

# Possible states of optimization problems

An optimization problem may have the following states:

- ▶ Infeasible (max problems,  $Z = -\infty$  and min problems,  $Z = +\infty$ )
- ► Feasible, optimal value finite but not attainable
- ► Feasible, optimal value finite and attainable
- Feasible, but optimal value is unbounded (max problems,  $Z=+\infty$  and min problems,  $Z=-\infty$ )

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# Pointwise maximum/minimum, no problem!

How to linearize functions such as  $\max_i \{\mathbf{a}_i^T x + b_i\}$  &  $\min_i \{\mathbf{a}_i^T x + b_i\}$ ?

▶ Define 
$$y = \max_{i} \{\mathbf{a}_{i}^{T}\mathbf{x} + b_{i}\} \implies y \geq \mathbf{a}_{i}^{T}\mathbf{x} + b_{i}, \forall i$$

▶ Define 
$$y = \min \{\mathbf{a}_i^T \mathbf{x} + b_i\} \implies y \leq \mathbf{a}_i^T \mathbf{x} + b_i, \forall i$$

How about the following problem?

$$Z = \underset{\mathbf{x} \ge 0}{\mathsf{minimize}} \qquad \qquad \|\mathbf{x}\|_1 = \sum_i |x_i| \qquad \qquad \textbf{(3a)}$$

subject to 
$$A\mathbf{x} = \mathbf{b}$$
 (3b)

Note  $|x_i| = \max\{x_i, -x_i\}$ . Define  $y_i = |x_i|$ 

$$Z = \underset{\mathbf{x} \ge 0, \mathbf{y}}{\mathsf{minimize}} \qquad \qquad \sum_{i} y_{i} \tag{4a}$$

subject to 
$$A\mathbf{x} = \mathbf{b}$$
 (4b)

$$y_i \ge x_i, \forall i$$
 (4c)

$$y_i \ge -x_i, \forall i$$
 (4d)

# Pointwise maximum/minimum, no problem!

#### How about the following problem?

$$Z = \underset{\mathbf{x} \geq 0}{\operatorname{minimize}} \qquad \qquad \|\mathbf{x}\|_{\infty} = \max_{i} \{|x_i|\}$$
 (5a)

subject to 
$$Ax = b$$
 (5b)

▶ Define  $y = \max_i\{|x_i|\}$ 

$$Z = \underset{\mathbf{x} \ge 0, y}{\mathsf{minimize}} \qquad \qquad y \tag{6a}$$

subject to 
$$A\mathbf{x} = \mathbf{b}$$
 (6b)

$$y \ge x_i, \forall i$$
 (6c)

$$y \ge -x_i, \forall i \tag{6d}$$

# **Linear Integer Program**

$$Z = \underset{\mathbf{x}}{\text{minimize}} \qquad \mathbf{c}^{T}\mathbf{x} \qquad (7a)$$
subject to 
$$A\mathbf{x} = \mathbf{b} \qquad (7b)$$

$$x_{i} \in \mathbb{Z}_{+}, i = 1, ..., p \qquad (7c)$$

$$x_{i} \in \mathbb{R}_{+}, i = p + 1, ..., n \qquad (7d)$$

- ► Generally, solving IP is more difficult than solving an LP. We use various tools from LP to approach this difficult problem.
- ▶ Better formulating the problem makes a lot of difference.

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(Fleet sizing Problem) CEGE 5214 A transit agency is going to optimize its fleet size and type to maximize its revenue. Possible vehicle types are:

- ▶ Vans, capacity 6, purchase cost \$20, projected revenue \$96
- Regular buses, capacity 28, purchase cost \$120, projected revenue \$400
- Articulated buses, capacity 56, purchase cost \$220, projected revenue \$900

#### Constraints:

- Available budget is \$2,000
- ► The agency has 25 drivers who have 20% vacation/sick/no-show rate
- ► The fleet should provide a minimum capacity of 450
- At least 30% of the fleet should be vans for demand-responsive service
- ► At least 10 regular buses are needed for the fixed routes
- Exactly 2 articulated buses are needed for an express route

(Shortest path problem) Given a directed graph G(N,A), cost of traversing links  $c:A\mapsto \mathbb{R}$ , find the shortest path from  $s\in N$  to  $t\in N$ .

(Maximum flow problem) Given a directed graph G(N,A), cost of traversing links  $c:A\mapsto \mathbb{R}$ , and capacities of links  $u:A\mapsto \mathbb{R}$ , find the maximum flow possible to send from  $s\in N$  to  $t\in N$ .

(Transportation Problem) We have n factories each supplying  $a_i$  units of construction lumber and m cities each with  $b_i$  demand of lumber. If the transportation cost of each unit from factory i to city j is  $c_{ij}$ , formulate a program that minimizes the total transportation cost while serving the demand in all the cities.

# Thank you!