

# 问求算法设计与实践06

线性规划建模

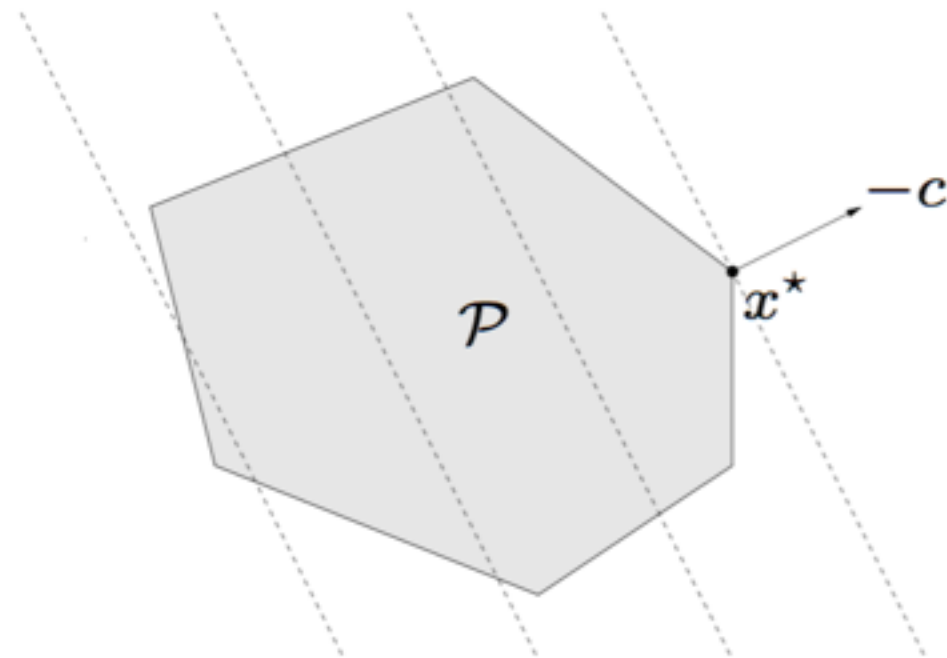
# Linear Programming

- 定义
- 算法
  - Simplex Algorithm(1947)
  - Interior Point Algorithm(1984),  $O(N^{3.5} \cdot L)$
  - 软件包: linprog, lingo, cvxopt...

# Linear program (LP)

$$\begin{array}{ll}\text{minimize} & c^T x + d \\ \text{subject to} & Gx \preceq h \\ & Ax = b\end{array}$$

- convex problem with affine objective and constraint functions
- feasible set is a polyhedron



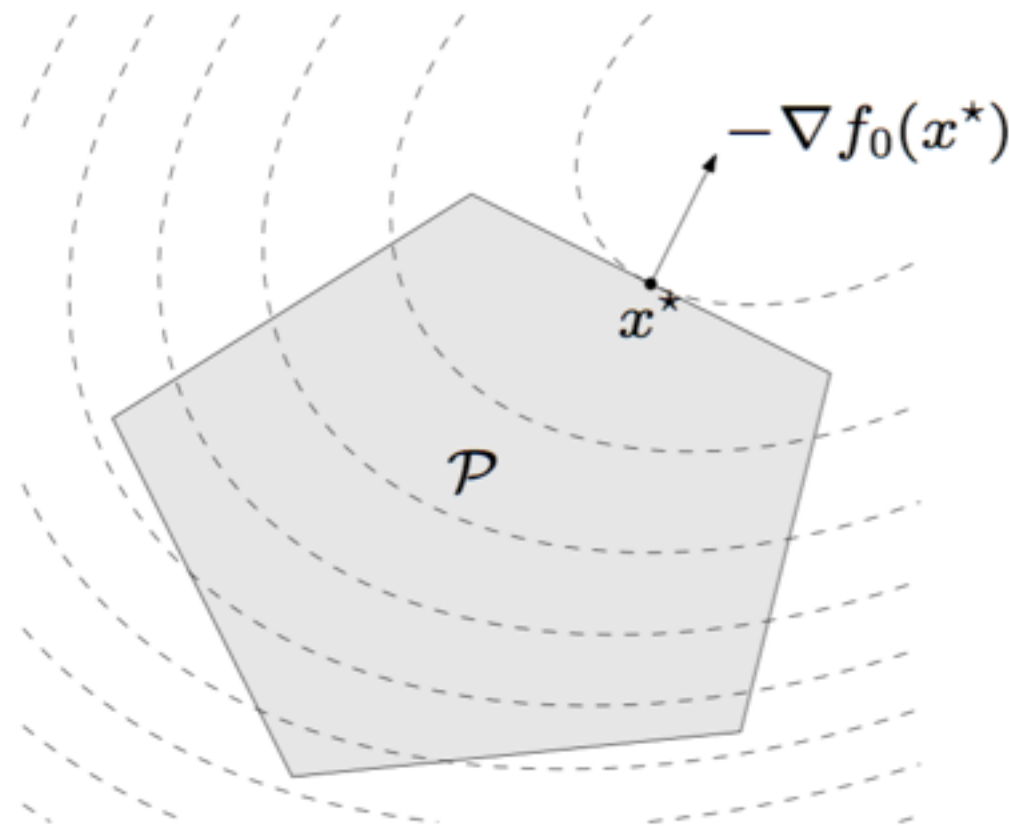
# More Programmings

- Linear-fractional Programming
- Quadratic Programming (with Quadratic Constraints)
- Geometric Programming
- Semidefinite Programming
- ...
- All about Convex!
- hint:dual problem

## Quadratic program (QP)

$$\begin{array}{ll}\text{minimize} & (1/2)x^T P x + q^T x + r \\ \text{subject to} & Gx \preceq h \\ & Ax = b\end{array}$$

- $P \in \mathbf{S}_{+}^n$ , so objective is convex quadratic
- minimize a convex quadratic function over a polyhedron



## Quadratically constrained quadratic program (QCQP)

$$\begin{array}{ll}\text{minimize} & (1/2)x^T P_0 x + q_0^T x + r_0 \\ \text{subject to} & (1/2)x^T P_i x + q_i^T x + r_i \leq 0, \quad i = 1, \dots, m \\ & Ax = b\end{array}$$

- $P_i \in \mathbf{S}_{+}^n$ ; objective and constraints are convex quadratic
- if  $P_1, \dots, P_m \in \mathbf{S}_{++}^n$ , feasible region is intersection of  $m$  ellipsoids and an affine set

## Two-way partitioning

$$\begin{array}{ll}\text{minimize} & x^T W x \\ \text{subject to} & x_i^2 = 1, \quad i = 1, \dots, n\end{array}$$

- a nonconvex problem; feasible set contains  $2^n$  discrete points
- interpretation: partition  $\{1, \dots, n\}$  in two sets;  $W_{ij}$  is cost of assigning  $i, j$  to the same set;  $-W_{ij}$  is cost of assigning to different sets

### dual function

$$\begin{aligned}g(\nu) &= \inf_x (x^T W x + \sum_i \nu_i (x_i^2 - 1)) = \inf_x x^T (W + \mathbf{diag}(\nu)) x - \mathbf{1}^T \nu \\ &= \begin{cases} -\mathbf{1}^T \nu & W + \mathbf{diag}(\nu) \succeq 0 \\ -\infty & \text{otherwise} \end{cases}\end{aligned}$$

**lower bound property:**  $p^* \geq -\mathbf{1}^T \nu$  if  $W + \mathbf{diag}(\nu) \succeq 0$

example:  $\nu = -\lambda_{\min}(W)\mathbf{1}$  gives bound  $p^* \geq n\lambda_{\min}(W)$

# Diet Problem

A healthy diet contains  $m$  different nutrients in quantities at least equal to  $b_1, \dots, b_m$ . We can compose such a diet by choosing nonnegative quantities  $x_1, \dots, x_n$  of  $n$  different foods. One unit quantity of food  $j$  contains an amount  $a_{ij}$  of nutrient  $i$ , and has a cost of  $c_j$ . We want to determine the cheapest diet that satisfies the nutritional requirements. This problem can be formulated as the LP

$$\begin{array}{ll}\text{minimize} & c^T x \\ \text{subject to} & Ax \succeq b \\ & x \succeq 0.\end{array}$$



# Shortest Path Problem

- $D_i$ 表示到第 $i$ 个点的最短路
- $w(u,v)$ 表示 $u \rightarrow v$ 的边权
- 求 $S \rightarrow T$ 的最短路

# Minimum-Cost Flow

- Consider a network of  $n$  nodes, with directed links connecting each pair of nodes. The variables in the problem are the flows on each link:  $X_{ij}$  will denote the flow from node  $i$  to node  $j$ . The cost of the flow along the link from node  $i$  to node  $j$  is given by  $C_{ij} * X_{ij}$ , where  $C_{ij}$  are given constants. The total cost across the network is

$$C = \sum C_{ij} * X_{ij}$$

- Each link flow  $X_{ij}$  is also subject to a given lower bound  $L_{ij}$  (usually assumed to be nonnegative) and an upper bound  $U_{ij}$ .
- The problem is to minimize the total cost of flow through the network, subject to the constraints described above. Formulate this problem as an LP.

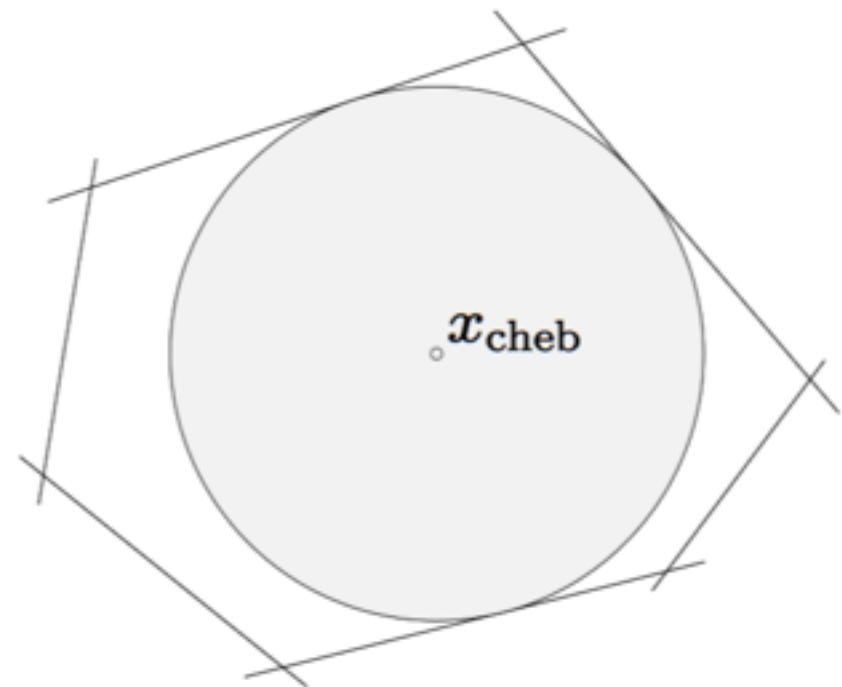
## Chebyshev center of a polyhedron

Chebyshev center of

$$\mathcal{P} = \{x \mid a_i^T x \leq b_i, \ i = 1, \dots, m\}$$

is center of largest inscribed ball

$$\mathcal{B} = \{x_c + u \mid \|u\|_2 \leq r\}$$



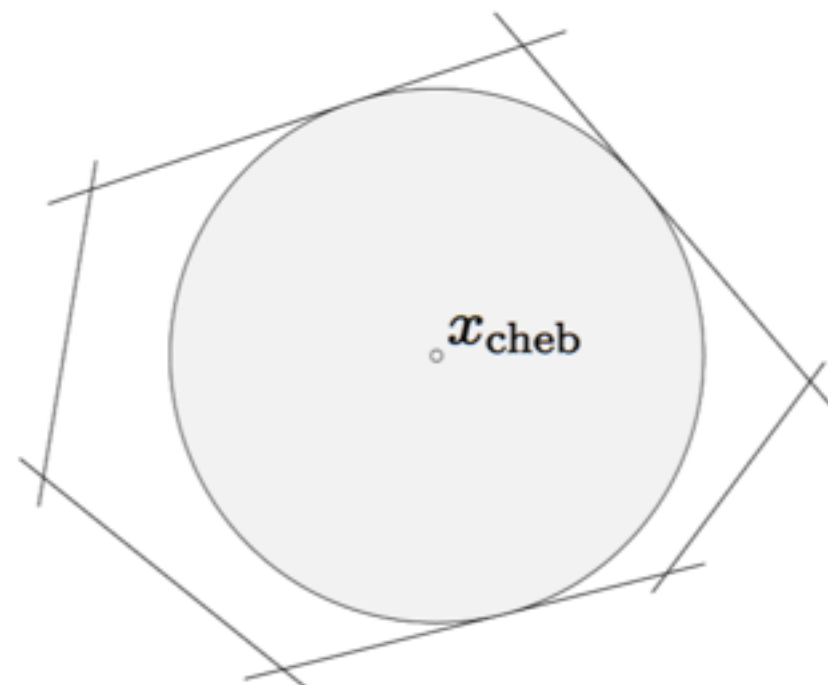
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- $a_i^T x \leq b_i$  for all  $x \in \mathcal{B}$  if and only if

$$\sup\{a_i^T (x_c + u) \mid \|u\|_2 \leq r\} = a_i^T x_c + r\|a_i\|_2 \leq b_i$$

- hence,  $x_c, r$  can be determined by solving the LP

$$\begin{array}{ll} \text{maximize} & r \\ \text{subject to} & a_i^T x_c + r\|a_i\|_2 \leq b_i, \quad i = 1, \dots, m \end{array}$$

# Dynamic activity planning

We consider the problem of choosing, or planning, the activity levels of  $n$  activities, or sectors of an economy, over  $N$  time periods. We let  $x_j(t) \geq 0$ ,  $t = 1, \dots, N$ , denote the activity level of sector  $j$ , in period  $t$ . The activities both consume and produce products or goods in proportion to their activity levels. The amount of good  $i$  produced per unit of activity  $j$  is given by  $a_{ij}$ . Similarly, the amount of good  $i$  consumed per unit of activity  $j$  is  $b_{ij}$ . The total amount of goods produced in period  $t$  is given by  $Ax(t) \in \mathbf{R}^m$ , and the amount of goods consumed is  $Bx(t) \in \mathbf{R}^m$ . (Although we refer to these products as ‘goods’, they can also include unwanted products such as pollutants.)

The goods consumed in a period cannot exceed those produced in the previous period: we must have  $Bx(t+1) \preceq Ax(t)$  for  $t = 1, \dots, N$ . A vector  $g_0 \in \mathbf{R}^m$  of initial goods is given, which constrains the first period activity levels:  $Bx(1) \preceq g_0$ . The (vectors of) excess goods not consumed by the activities are given by

$$\begin{aligned} s(0) &= g_0 - Bx(1) \\ s(t) &= Ax(t) - Bx(t+1), \quad t = 1, \dots, N-1 \\ s(N) &= Ax(N). \end{aligned}$$



The objective is to maximize a discounted total value of excess goods:

$$c^T s(0) + \gamma c^T s(1) + \cdots + \gamma^N c^T s(N),$$

where  $c \in \mathbf{R}^m$  gives the values of the goods, and  $\gamma > 0$  is a discount factor. (The value  $c_i$  is negative if the  $i$ th product is unwanted, *e.g.*, a pollutant;  $|c_i|$  is then the cost of disposal per unit.)

Putting it all together we arrive at the LP

$$\begin{array}{ll} \text{maximize} & c^T s(0) + \gamma c^T s(1) + \cdots + \gamma^N c^T s(N) \\ \text{subject to} & x(t) \succeq 0, \quad t = 1, \dots, N \\ & s(t) \succeq 0, \quad t = 0, \dots, N \\ & s(0) = g_0 - Bx(1) \\ & s(t) = Ax(t) - Bx(t+1), \quad t = 1, \dots, N-1 \\ & s(N) = Ax(N), \end{array}$$

with variables  $x(1), \dots, x(N), s(0), \dots, s(N)$ . This problem is a standard form LP; the variables  $s(t)$  are the slack variables associated with the constraints  $Bx(t+1) \preceq Ax(t)$ .

# Reference

[1] Boyd, Stephen, and Lieven Vandenberghe.  
*Convex optimization*. Cambridge university press,  
2004.

[2][http://stanford.edu/class/ee364a/lectures/  
problems.pdf](http://stanford.edu/class/ee364a/lectures/problems.pdf)