

Linear programs

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

Equational form

$$\begin{aligned} &\text{maximize } c^T x \\ &\text{subject to } Ax = b \leftarrow \text{equalities} \\ &\quad x \geq 0 \leftarrow \text{the only inequalities} \end{aligned}$$

Transforming an arbitrary linear program to equational form

- Inequalities to equalities

$$2x_1 - x_2 \leq 4 \rightsquigarrow 2x_1 - x_2 + \underbrace{z_1}_{\text{slack variable}} = 4$$

$$z_1 \geq 0$$

$$x_1 + 3x_2 \geq 4 \rightsquigarrow x_1 + 3x_2 - \underbrace{z_2}_{\text{slack variable}} = 4$$

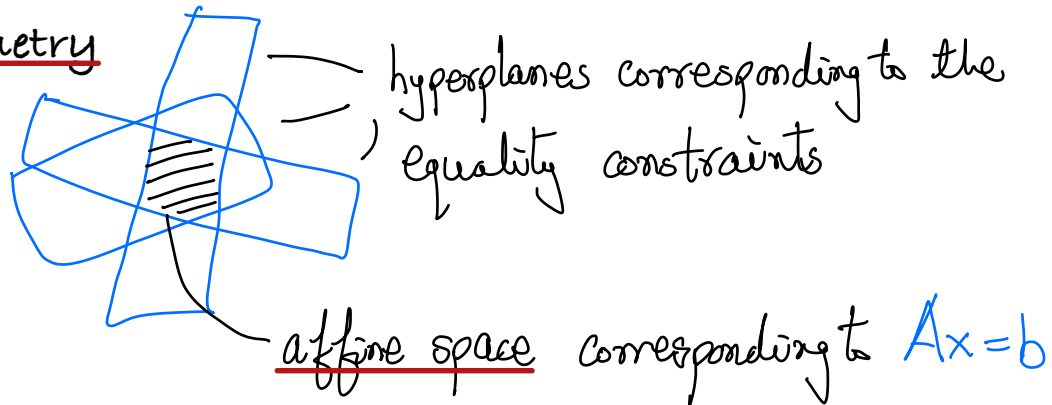
$$z_2 \geq 0$$

- Unconstrained variables

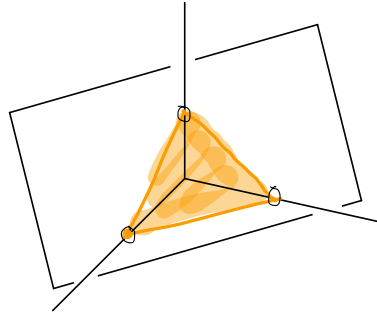
For each variable x_i introduce two new variables y_i and z_i . Replace x_i by $y_i - z_i$ everywhere, and add the non-negativity constraints

$$y_i \geq 0 \quad \text{and} \quad z_i \geq 0$$

Geometry



The feasible region is the intersection of this affine space with the positive orthant corresponding to $x \geq 0$

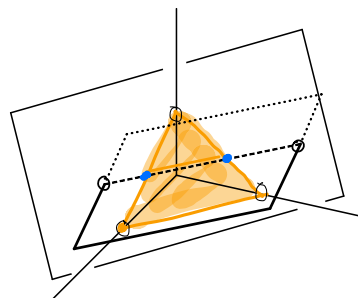
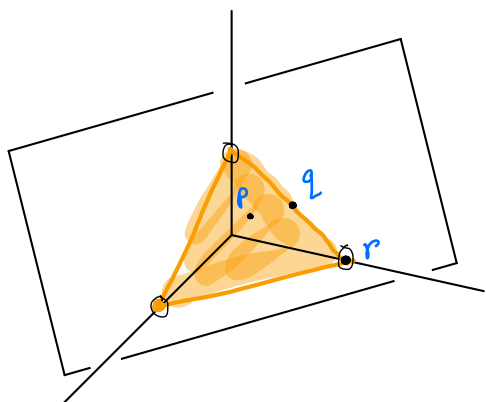


Assumptions

- The system $Ax=b$ has at least one solution;
- The rows of the matrix A are linearly independent

A is an $m \times n$ matrix with m linearly independent rows.

Basic feasible solutions



For an $m \times n$ matrix A and a subset B of $[n] = \{1, 2, \dots, n\}$ let A_B denote the matrix consisting of those columns of A whose indices appear in B .

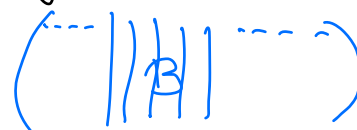
A basic feasible solution of the linear program

$$\begin{aligned} & \text{maximize } c^T x \\ & \text{subject to } Ax = b \quad (A \in \mathbb{R}^{m \times n}) \\ & \quad \quad \quad x \geq 0 \end{aligned}$$



is a feasible solution $x \in \mathbb{R}^n$ for which there exists an m -element set B such that

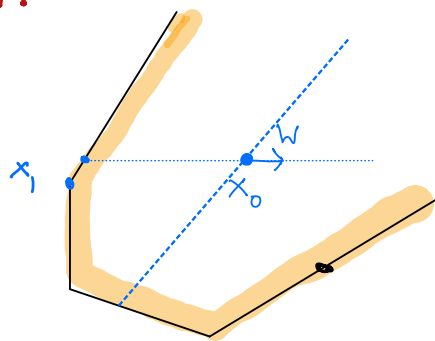
- columns of A_B are linearly independent;
- $x_j = 0$ for $j \notin B$.



Proposition: For every $B \subseteq \{1, 2, \dots, n\}$ with A_B non-singular, there exists at most one feasible solution $x \in \mathbb{R}^n$ with $x_j = 0$ for all $j \notin B$. $Bx_B = b$

Theorem: If the objective function of an LP is bounded above, then for every feasible solution x_0 , there is a basic feasible solution \tilde{x} , s.t.

Presented again in Lecture 4. $\tilde{C}^T \tilde{x} \geq \tilde{C}^T x_0$.



$$Ax_0 = b$$

$$K = \{j : x_0[j] > 0\}$$

If A_K is singular,

there is a $w \neq 0$ supported on K s.t. $Aw = 0$.

Corollary:

LP feasible and bounded



\exists optimum solution

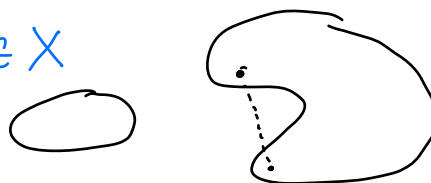


\exists basic feasible optimum solution

Convexity and convex polyhedra

$X \subseteq \mathbb{R}^n$ is convex if $\forall u, v \in X \quad \forall t \in [0, 1]$

$$tu + (1-t)v \in X$$



$S \subseteq \mathbb{R}^n$: The convex hull of S is the smallest convex set containing S , that is, the intersection of all convex sets that contain S .

A convex combination of points $x_1, x_2, \dots, x_k \in \mathbb{R}^n$ is a point x that can be expressed as

$$x = t_1 x_1 + t_2 x_2 + \dots + t_k x_k \text{ where } t_i \geq 0 \\ \text{and } t_1 + t_2 + \dots + t_k = 1.$$

Proposition: Let $X \subseteq \mathbb{R}^n$. Then,

Convex hull of $X = \left\{ x \in \mathbb{R}^n : x \text{ is a convex combination of some finite set of points in } X \right\}.$

$$\text{hyperplane} = \left\{ x \in \mathbb{R}^n : a^T x = b \right\}, \quad \begin{array}{l} a \neq 0, \\ \uparrow \\ \mathbb{R}^n \end{array}, \quad \begin{array}{l} b \\ \uparrow \\ \mathbb{R} \end{array}$$

$$\text{closed half-space} = \left\{ x \in \mathbb{R}^n : a^T x \geq b \right\}$$

A convex polyhedron is an intersection of finitely many closed half-spaces.

A bounded convex polyhedron is a convex polytope.

The dimension of a convex polyhedron $P \subseteq \mathbb{R}^n$ is the smallest dimension of an affine subspace containing P .

Examples

- Cubes
- Crosspolytope
- Simplex

vertices and basic feasible solutions

P : a polyhedron

A point v in P is a vertex of P if there is a nonzero vector c s.t.

$$c^T v > c^T y \quad \forall y \in P \setminus \{v\}.$$

The $H = \{x \in \mathbb{R}^n : c^T x = c^T v\}$ intersects P exactly at v , and P lies entirely in one of the closed half-spaces defined by H .

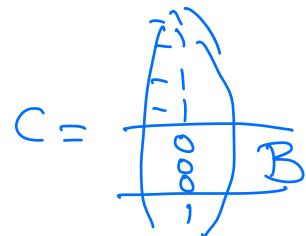
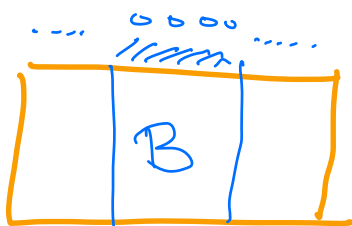
Theorem: P : the set of all feasible solutions of a linear program in equational form.

Convex polyhedron

(i) v is a vertex of P



(ii) v is a basic feasible solution of the LP.



Arbitrary linear programs

A basic feasible solution of a linear program

maximize $c^T x$

subject to $Ax \leq b$
 $=$ (not necessarily in equational form)

$A \in \mathbb{R}^{m \times n}$
 $b \in \mathbb{R}^m$
 $c \in \mathbb{R}^n$

is a basic feasible solution for which some n linearly independent constraints are satisfied with equality.

equational form

$$\left(\begin{array}{c} \overbrace{\hspace{2cm}}^n \\ \underbrace{\hspace{2cm}}_m \end{array} \begin{array}{c} \overbrace{\hspace{1cm}}^m \\ \underbrace{\hspace{1cm}}_m \end{array} B \right)$$

$$x \geq 0$$

arbitrary

$$\left(\begin{array}{c} \overbrace{\hspace{2cm}}^n \\ \underbrace{\hspace{2cm}}_{m+n} \end{array} \begin{array}{c} \overbrace{\hspace{1cm}}^m \\ \underbrace{\hspace{1cm}}_m \end{array} B \begin{array}{c} \\ \\ -1 \\ -1 \\ -1 \\ -1 \\ -1 \\ -1 \\ -1 \\ -1 \end{array} \right)$$

- There are (general) LPs none of whose infinitely many optimal solutions is basic.
- Vertices and extremal points
 - A point in the convex set where some linear function attains its unique maximum.
vertex
 - A point in the convex set that cannot be written as a convex combination of two other points in the convex set.
extreme point

For convex polyhedra x is a vertex
 \Updownarrow
 x is an extreme point

The main theorem of convex polytopes

A polytope is equal to the convex hull of its set of vertices.

Not obvious!