

## Lecture 4

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## 1 Introduction

Last time we talked about polyhedra and polytopes. This time we will define bounded polyhedra and discuss their relationship with polytopes. Recall from the last lecture the following definitions.

A polyhedron is  $P = \{x \in \mathbb{R}^n : Ax \leq b\}$ ,  $A \in \mathbb{R}^{m \times n}$ ,  $m \geq n$ .

A polytope is  $Q = \text{conv}(v_1, \dots, v_k)$  for finite  $k$ .

$x \in P$  is a vertex if  $\exists c \in \mathbb{R}^n$  such that  $c^T x < c^T y$  for all  $y \in P$ ,  $y \neq x$ .

$x \in P$  is an extreme point if  $\nexists y, z \in P$ ,  $y, z \neq x$  such that  $x = \lambda y + (1 - \lambda)z$ ,  $\lambda \in [0, 1]$ .

$x \in P$  is a basic feasible solution if  $x \in P$  and it is basic (i.e., the rank of  $A_{=}$  is  $n$ ).

Notice that the number of vertices of  $P$  is finite since given the  $m$  constraints in  $Ax \leq b$ , we can choose  $n$  of them to be met with equality; thus there are at most  $\binom{m}{n}$  basic solutions.

## 2 Polyhedra and Polytopes

Now we are interested in the following two questions:

- Q1: When is a polytope a polyhedron?
- A1: A polytope is always a polyhedron.
  
- Q2: When is a polyhedron a polytope?
- A2: A polyhedron is almost always a polytope.

We can give a counterexample to show why a polyhedron is not always but almost always a polytope: an unbounded polyhedron is not a polytope. See Figure 1.

**Definition 1** A polyhedron  $P$  is bounded if  $\exists M > 0$ , such that  $\|x\| \leq M$  for all  $x \in P$ .

What we can show is this: every bounded polyhedron is a polytope, and vice versa. In this lecture, we will show one side of the proof in one direction; we will show the other direction in the next lecture. To start with, we need the following lemma.

**Lemma 1** Any polyhedron  $P = \{x \in \mathbb{R}^n : Ax \leq b\}$  is convex.

**Proof:** If  $x, y \in P$ , then  $Ax \leq b$  and  $Ay \leq b$ . Therefore,

$$A(\lambda x + (1 - \lambda)y) = \lambda Ax + (1 - \lambda)Ay \leq \lambda b + (1 - \lambda)b = b.$$

Thus  $\lambda x + (1 - \lambda)y \in P$ . □

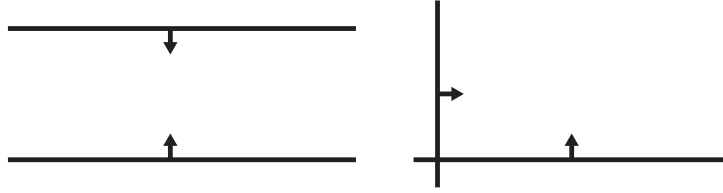


Figure 1: Examples of unbounded polyhedra that are not polytopes. (left) No extreme points, (right) one extreme point.

### 3 Representation of Bounded Polyhedra

We can now show the following theorem.

**Theorem 2 (Representation of Bounded Polyhedra)** *A bounded polyhedron  $P$  is the set of all convex combinations of its vertices, and is therefore a polytope.*

**Proof:** Let  $v_1, v_2, \dots, v_k$  be the vertices of  $P$ . Since  $v_i \in P$  and  $P$  is convex (by previous lemma), then any convex combination  $\sum_{i=1}^k \lambda_i v_i \in P$ . So it only remains to show that any  $x \in P$  can be written as  $x = \sum_{i=1}^k \lambda_i v_i$ , with  $\lambda_i \geq 0$  and  $\sum_{i=1}^k \lambda_i = 1$ .

Let  $A_ =$  be all the constraints that  $x$  meets with equality (all rows  $a_i$  such that  $a_i x = b_i$ ). Let  $ra(x)$  be the rank of the corresponding  $A_ =$ . Recall from last time that  $ra(x) = n$  if and only if  $x$  is a vertex of  $P$ . Now we prove the theorem by induction on  $n - ra(x)$ .

*Base case:* Let  $n - ra(x) = 0$ . Then  $ra(x) = n$  and since  $x \in P$ ,  $x$  is a basic feasible solution, and therefore a vertex of  $P$ .

*Inductive Step:* Suppose we have shown that for any  $y \in P$  such that  $n - ra(y) < \ell$  for some  $\ell > 0$ ,  $y$  can be written as a convex combination of  $v_1, v_2, \dots, v_k$ . Consider  $x \in P$  with  $ra(x) = n - \ell < n$ . Then the rank of  $A_ = < n$ , and thus there exists  $z$  such that  $A_ = z = 0$ . Since  $P$  is bounded, there exist constants  $\bar{\alpha} > 0$  and  $\underline{\alpha} < 0$  such that  $x + \alpha z \in P$  if and only if  $\underline{\alpha} \leq \alpha \leq \bar{\alpha}$ . Geometrically, this is equivalent to moving from  $x$  in the direction  $\alpha z$  until we run into a constraint.

Then we can express  $x$  as

$$x = \frac{\bar{\alpha}}{\bar{\alpha} - \underline{\alpha}}(x + \underline{\alpha}z) + \frac{-\underline{\alpha}}{\bar{\alpha} - \underline{\alpha}}(x + \bar{\alpha}z).$$

Therefore,  $x$  is a convex combinations of two points in  $P$ . Now all we need to show is that  $x + \underline{\alpha}z$  and  $x + \bar{\alpha}z$  are convex combinations of vertices. Since  $x + \bar{\alpha}z \in P$ , but  $x + \alpha z \notin P$  for  $\alpha > \bar{\alpha}$ , there exists some constraint  $a_j$  such that  $a_j x < b_j$ , but  $a_j(x + \bar{\alpha}z) = b_j$ . This implies that  $ra(x + \bar{\alpha}z) > ra(x)$ , so then  $n - ra(x + \bar{\alpha}z) < n - ra(x) = \ell$ . Therefore,  $x + \bar{\alpha}z$  can be expressed as a convex combination of vertices  $v_1, v_2, \dots, v_k$  by induction; we suppose  $x + \bar{\alpha}z = \sum_{i=1}^k \alpha_i v_i$ , where  $\alpha_i \geq 0$  and  $\sum_{i=1}^k \alpha_i = 1$ . Similarly, it must be the case that  $x + \underline{\alpha}z$  is a convex combination of the vertices, and we can write  $x + \underline{\alpha}z = \sum_{i=1}^k \beta_i v_i$ , where  $\beta_i \geq 0$  and  $\sum_{i=1}^k \beta_i = 1$ .

Therefore, we have

$$\begin{aligned}
x &= \frac{\bar{\alpha}}{\bar{\alpha} - \underline{\alpha}}(x + \underline{\alpha}z) + \frac{-\underline{\alpha}}{\bar{\alpha} - \underline{\alpha}}(x + \bar{\alpha}z) \\
&= \frac{\bar{\alpha}}{\bar{\alpha} - \underline{\alpha}} \sum_{i=1}^k \alpha_i v_i + \frac{-\underline{\alpha}}{\bar{\alpha} - \underline{\alpha}} \sum_{i=1}^k \beta_i v_i \\
&= \sum_{i=1}^k \left( \frac{\bar{\alpha}}{\bar{\alpha} - \underline{\alpha}} \alpha_i + \frac{-\underline{\alpha}}{\bar{\alpha} - \underline{\alpha}} \beta_i \right) v_i \\
&= \sum_{i=1}^k \delta_i v_i,
\end{aligned}$$

where  $\delta_i = \frac{\bar{\alpha}}{\bar{\alpha} - \underline{\alpha}} \alpha_i + \frac{-\underline{\alpha}}{\bar{\alpha} - \underline{\alpha}} \beta_i \geq 0$  and

$$\begin{aligned}
\sum_{i=1}^k \delta_i &= \sum_{i=1}^k \left( \frac{\bar{\alpha}}{\bar{\alpha} - \underline{\alpha}} \alpha_i + \frac{-\underline{\alpha}}{\bar{\alpha} - \underline{\alpha}} \beta_i \right) \\
&= \frac{\bar{\alpha}}{\bar{\alpha} - \underline{\alpha}} \sum_{i=1}^k \alpha_i + \frac{-\underline{\alpha}}{\bar{\alpha} - \underline{\alpha}} \sum_{i=1}^k \beta_i \\
&= \frac{\bar{\alpha}}{\bar{\alpha} - \underline{\alpha}} + \frac{-\underline{\alpha}}{\bar{\alpha} - \underline{\alpha}} = 1.
\end{aligned}$$

Thus  $x$  is a convex combination of the vertices. □

## 4 Separating Hyperplane Theorem

To begin showing the proof in the opposite direction (that is, showing that every polytope is a bounded polyhedron), we will need a theorem called the *separating hyperplane theorem*. To prove the theorem, we will use the following theorem from analysis, which we give without proof.

**Theorem 3 (Weierstrass)** *Let  $C \subseteq \mathfrak{R}^n$  be a closed, non-empty and bounded set. Let  $f : C \rightarrow \mathfrak{R}$  be continuous on  $C$ . Then  $f$  attains a maximum (and a minimum) on some point of  $C$ .*

Suppose  $f(x) = \frac{1}{2}\|x - y\|$ , for all  $x \in C$ . We'd like to apply Weierstrass' theorem to find the minimizer of  $f$  in  $C$ , but  $C$  may not be bounded. To get around this, we pick some  $q \in C$ , which we can do since  $C$  is non-empty. Then, let  $\hat{C} = \{x \in C : \|q - y\| \geq \|x - y\|\}$ .  $\hat{C}$  is closed, non-empty and bounded; we see that  $\hat{C}$  is bounded since for  $x \in \hat{C}$ , we have  $\|x\| \leq \|y\| + \|y - x\|$  by the triangle inequality and  $\|y\| + \|y - x\| \leq \|y\| + \|q - y\|$  by the definition of  $\hat{C}$ ; both  $\|y\|$  and  $\|q - y\|$  are constant terms. Now we can apply Weierstrass' theorem on  $\hat{C}$  to find a point  $z$  that minimizes  $f$ .

**Theorem 4 (Separating Hyperplane)** *Let  $C \subseteq \mathfrak{R}^n$  be closed, non-empty and convex set. Let  $y \notin C$ , then there exists a hyperplane  $a \neq 0$ ,  $a \in \mathfrak{R}^n$ ,  $b \in \mathfrak{R}$ , such that  $a^T y > b$  and  $a^T x < b$ , for all  $x \in C$ .*

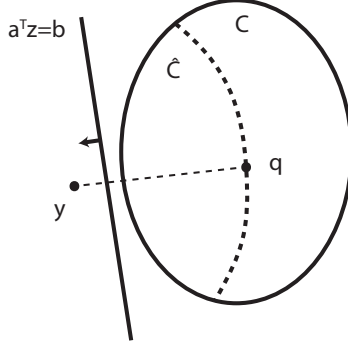


Figure 2: Separating hyperplane

**Proof:** Define

$$f(x) = \frac{1}{2} \|x - y\|^2$$

$$\hat{C} = \{x \in C : \|q - y\| \geq \|q - x\|\}.$$

Apply Weierstrass' theorem. Let  $z$  be the minimizer of  $f$  in  $\hat{C}$ . Note that for any  $x \in C - \hat{C}$ ,  $f(z) \leq f(q) < f(x)$ , and therefore  $z$  minimizes  $f$  over all of  $C$ , since any  $x \notin \hat{C}$  must have been further away from  $y$  than  $q$ .

Let  $a = y - z$ . Then  $a \neq 0$ , since  $z \in C, y \notin C$ . Let  $b = \frac{1}{2}(a^T y + a^T z)$ . Then,

$$0 < a^T a = a^T (y - z) = a^T y - a^T z$$

so then

$$a^T y > a^T z \Rightarrow 2a^T y > a^T y + a^T z \Rightarrow a^T y > \frac{1}{2}(a^T y + a^T z) = b.$$

It remains to show that  $a^T x < b$  for all  $x \in C$ . Let  $x_\lambda = (1 - \lambda)z + \lambda x \in C$  for  $0 < \lambda \leq 1$ . Since  $z$  minimizes  $f$  over  $C$ ,  $f(z) \leq f(x_\lambda)$ . Thus,

$$\begin{aligned} f(x_\lambda) &= \frac{1}{2}((1 - \lambda)z + \lambda x - y)^T((1 - \lambda)z + \lambda x - y) = \frac{1}{2}(z - y + \lambda(x - z))^T(z - y + \lambda(x - z)) \\ &\geq \frac{1}{2}(z - y)^T(z - y) = f(z). \end{aligned}$$

Rewriting, we obtain

$$\begin{aligned} \frac{1}{2}[2(z - y)^T \lambda(x - z) + \lambda^2(x - z)^T(x - z)] &\geq 0 \\ (z - y)^T(x - z) + \frac{1}{2}\lambda(x - z)^T(x - z) &\geq 0 \\ a^T(z - x) + \frac{1}{2}\lambda(x - z)^T(x - z) &\geq 0 \end{aligned}$$

or

$$a^T(z - x) \geq -\frac{1}{2}\lambda(x - z)^T(x - z).$$

But we can take  $\lambda \rightarrow 0$  arbitrarily small, so  $a^T(z - x) \geq 0$  which implies  $a^T z \geq a^T x$ . Using the fact that  $a^T z < a^T y$ ,

$$b = \frac{1}{2}(a^T y + a^T z) \geq \frac{1}{2}(2a^T z) = a^T z > a^T x.$$

□