#### **ORIE 6300** Mathematical Programming I

September 6, 2016

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Lecture 5

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# 1 Review

A while back, we defined polyhedrons and polytopes as follows.

**Definition 1** A Polyhedron is  $P = \{x \in \mathbb{R}^n : Ax \leq b\}$ 

**Definition 2** A Polytope is given by  $Q = conv(v_1, v_2, ..., v_k)$ , where the  $v_i$  are the vertices of the polytope, for k finite.

Also recall the equivalence of extreme points, vertices and basic feasible solutions, and recall the definition of a bounded polyhedron.

**Definition 3** A polyhedron P is bounded iff  $\exists M > 0$  such that  $||x|| \leq M$ ,  $\forall x \in P$ .

We showed bounded polyhedra were polytopes by taking the extreme points and seeing that they were the vertices for P as a polytope.

Recall also the Separating Hyperplane Theorem from a previous lecture.

**Theorem 1** (Separating Hyperplane) Let  $C \subseteq \mathbb{R}^n$  be a closed, nonempty and convex set. Let  $y \in \mathbb{R}^n \setminus C$  and let

$$x^* = P_C(y) := \operatorname{argmin}_{x \in C} \frac{1}{2} ||x - y||^2.$$

Then there exists a number  $b \in \mathbb{R}$ , such that with  $a = y - x^*$ , we have

$$(\forall x \in C) \qquad a^T x \le a^T x^* < b < a^T y.$$

### 2 The polar of a set

Now we want to prove that polytopes are bounded polyhedra. To do this, we need to introduce one more concept.

**Definition 4** Let  $S \subseteq \mathbb{R}^n$ . We call the set

$$S^{\circ} = \{ z \in \mathbb{R}^n : z^T x \le 1, \, \forall x \in S \},\$$

the polar set of S.

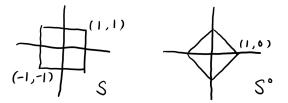


Figure 1: The polar of the  $\ell_{\infty}$  ball is the  $\ell_1$  norm ball.

**Example 1 (Polars of the**  $\ell_{\infty}$  and  $\ell_1$  balls.) Consider  $S = \{(x_1, x_2) : -1 \leq x_1 \leq 1, -1 \leq x_2 \leq 1\}$ , the region is shown in the left of Figure 1.

By definition,  $x \in S^{\circ}$  if, and only if,  $|x_1| + |x_2| = \sup_{(z_1, z_2) \in S} \{x_1 z_1 + x_2 z_2\} \leq 1$ . Thus,  $S^{\circ} = \{x \in \mathbb{R}^n : |x_1| + |x_2| \leq 1\}$ , which is shown on the right hand side of Figure 1.

Now let's consider  $S^{\circ\circ}$ . By definition,  $x \in (S^{\circ})^{\circ}$  if, and only if,  $\max\{|z_1|, |z_2|\} = \sup_{|x_1|+|x_2|\leq 1}\{x_1z_1+x_2z_2\} \leq 1$ . Thus,  $(S^{\circ})^{\circ} = \{(x_1, x_2): -1 \leq x_1 \leq 1, -1 \leq x_2 \leq 1\} = S$ .

**Lemma 2** If C is a closed convex subset of  $\mathbb{R}^n$  with  $0 \in C$ , then  $C^{\circ\circ} := (C^{\circ})^{\circ} = C$ .

### **Proof:**

- ( $\supseteq$ ) Suppose that  $x \in C$ . Then  $x \in C^{\circ\circ}$  if, and only if,  $z^T x \leq 1$  for all  $z \in C^{\circ}$ . This is clearly true because  $z \in C^{\circ}$  implies that  $z^T x \leq 1$ .
- ( $\subseteq$ ) We will show that if  $x \notin C$ , then  $x \notin C^{\circ\circ}$ . First note that C is closed and convex with at least  $z = 0 \in C$ . If  $x \notin C$ , then by the Separating Hyperplane Theorem, there exists  $0 \neq a \in \mathbb{R}^n$  and  $b \in \mathbb{R}$  with  $a^T x > b > a^T z$  for all  $z \in C$ . Since  $0 \in C$ , we have b > 0. Let  $\tilde{a} = a/b \neq 0$ . Therefore  $\tilde{a}^T x > 1 > \tilde{a}^T z$ , for all  $z \in C$ . This implies  $\tilde{a} \in C^{\circ}$ . But  $\tilde{a}^T x > 1$ , so  $x \notin C^{\circ\circ}$ .

Therefore  $C^{\circ\circ} = C$ .

## 3 Polytopes are Bounded Polyhedra

Now we can prove our result, at least sort of. We'll assume that 0 is in the interior of the polytope. We claim that this can be done without loss of generality; this is because we can translate the polytope to have  $0 \in P$ , apply the following proof and then translate back if needed.

**Theorem 3** If  $Q \subseteq \mathbb{R}^n$  is a polytope with 0 in the interior of Q, then Q is a (bounded) polyhedron.

**Proof:** Our proof strategy is as follows. (1) We will first show that the polar of a polytope is a polyhedron. (2) We then show that that since the polytope has 0 in its interior, then the polar of the polytope is bounded. So then  $P = Q^{\circ}$  is a bounded polyhedron. (3) We know from a previous lecture that any bounded polyhedron is a polytope, so  $P = Q^{\circ}$  is a polytope. (4) But then applying the proof that the polar of a polytope is a polyhedron, we get that  $P^{\circ} = Q^{\circ \circ} = Q$  (by the lemma above) is a polyhedron. It is easy to prove that a polytope is bounded. Thus, we must prove (1) and (2).

(1) We first prove that the polar of Q is a polyhedron. Let  $P = Q^{\circ}$ . Then we know that  $P^{\circ} = Q^{\circ \circ} = Q$ . Since Q is a polytope,  $Q = \operatorname{conv}\{v_1, \ldots, v_k\}$  for some k finite vectors  $v_1, \ldots, v_k \in \mathbb{R}^n$ . We claim that

$$P = \{ z \in \mathbb{R}^n : v_i^T z \le 1, \, i = 1, \dots, k \}.$$

One the one hand,  $P = Q^{\circ} = \{z \in \mathbb{R}^n : x^T z \leq 1, \forall x \in Q\}$ , so  $v_i^T z = z^T v_i \leq 1$  for i = 1, 2, ..., k, which implies that  $P \subseteq \{z \in \mathbb{R}^n : v_i^T z \leq 1, i = 1, ..., k\}$ . On the other hand, if  $z^T v_i \leq 1$  for i = 1, ..., k, then for any  $x \in Q$ , with  $x = \sum_{i=1}^k \lambda_i v_i$  for some  $\lambda_i \geq 0, \sum_i \lambda_i = 1$ , we have

$$z^T x = z^T \sum_{i=1}^k \lambda_i v_i = \sum_{i=1}^k \lambda_i (z^T v_i) \le \sum_{i=1}^k \lambda_i = 1,$$

which proves the claim. Thus P is a polyhedron.

(2) Because  $0 \in int(Q)$ , there exists some  $\epsilon > 0$  such that all  $x \in \mathbb{R}^n$  with  $||x|| < \epsilon$  lie in Q. If  $z \in P, z \neq 0$ , then, because  $||x|| < \epsilon$ , we have

$$x=\frac{\epsilon}{2}\frac{z}{||z||}\in Q.$$

Because  $P = Q^{\circ}$ ,

$$x^T z \le 1 \quad \Rightarrow \quad \frac{\epsilon z^T z}{2||z||} \le 1 \quad \Rightarrow \quad ||z|| \le \frac{2}{\epsilon}.$$

Hence P is a bounded polyhedron.

### 4 Normal Cone

Modern optimization theory crucially relies on a concept called the *normal cone*.

**Definition 5** Let  $S \subset \mathbb{R}^n$  be a closed, convex set. The normal cone of S is the set-valued mapping  $N_S : \mathbb{R}^n \to 2^{\mathbb{R}^n}$ , given by

$$N_S(x) = \begin{cases} \{g \in \mathbb{R}^n | (\forall z \in S) \ g^T(z - x) \le 0\} & \text{if } x \in S \\ \emptyset & \text{if } x \notin S \end{cases}$$

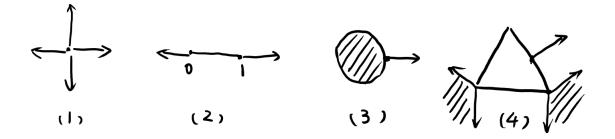


Figure 2: Normal cones of several convex sets.

**Example 2** We compute several normal cones; see Figure 2.

- 1. Let  $S = \{z\}$ .  $N_S(x) = \begin{cases} \mathbb{R}^n & \text{if } x = z \\ \emptyset & \text{otherwise} \end{cases}$
- 2. Let S = [0, 1].

$$N_S(x) = \begin{cases} \mathbb{R}_{\leq 0} & \text{if } x = 0\\ \mathbb{R}_{\geq 0} & \text{if } x = 1\\ \{0\} & \text{if } x \in (0, 1)\\ \emptyset & \text{otherwise} \end{cases}$$

3. Let  $S = \{x \mid ||x|| \le 1, x \in \mathbb{R}^n\}.$ 

$$N_S(x) = \begin{cases} \mathbb{R}_{\geq 0} x & if \|x\| = 1\\ \{0\} & if \|x\| < 1\\ \emptyset & otherwise \end{cases}$$

4. The normal cone of a triangle, computed at some but not all points, is depicted in Figure 2.

Normal cones satisfy several useful properties.

**Proposition 4** Let  $S \subseteq \mathbb{R}^n$  be a nonempty, closed, convex set. Then the following hold:

1. If  $x \in S$ , then  $N_S(x)$  is a convex cone, i.e.

$$(\forall \lambda_1 \ge 0), (\forall \lambda_2 \ge 0), (\forall g_1 \in N_S(x)), (\forall g_2 \in N_S(x)) \ \lambda_1 g_1 + \lambda_2 g_2 \in N_S(x).$$

- 2. Let  $y \in \mathbb{R}^n \setminus S$ , then  $P_S(y) = x \iff y x \in N_S(x)$ .
- 3. If  $x \in int(S)$ , then  $N_S(x) = \{0\}$ .
- 4. If  $x \in S$  and  $x \notin int(S)$ , then  $N_S(x) \supseteq \{0\}$ .

#### **Proof:**

1. We leave the proof of part 1 as an exercise.

2. ( $\Rightarrow$ ): By separating hyperplane theorem, with  $a = y - P_S(y) = y - x, \exists b \in \mathbb{R}, s.t.$ 

$$\begin{array}{ll} (\forall z \in S) & a^T z \leq a^T P_S(y) \leq a^T y \\ \Rightarrow & a^T (z - P_S(y)) \leq 0 \\ \Rightarrow & a \in N_S(P_S(y)) \end{array}$$

 $(\Leftarrow)$ : If  $y - x \in N_S(x)$ , then

$$\begin{aligned} \forall z \in S & (y-x)^T (z-x) \leq 0 \\ \Leftrightarrow & (y-x)^T (y-x) + (y-x)^T (z-y) \leq 0 \\ \Rightarrow & ||y-x||^2 \leq (y-x)^T (y-z) \leq ||y-x|| \ ||y-z|| \ (Cauchy - Schwarz \ inequality) \\ \Rightarrow & ||y-x|| \leq ||y-x|| \ \leq ||y-z||, \ \forall z \in S \\ \Rightarrow & x = P_S(y) \end{aligned}$$

3. Suppose that  $g \in N_S(x)$ . Because  $x \in int(S)$ , there exists  $\epsilon > 0$ , such that  $x + \epsilon g \in S$ . Therefore, we have

$$g^{T}((x + \epsilon g) - x) \le 0$$
  
$$\Rightarrow \qquad \epsilon g^{T}g \le 0$$
  
$$\Rightarrow \qquad g = 0.$$

Hence,  $N_S(x) = \{0\}.$ 

4. Since  $x \notin int(S)$ , there exists a sequence  $y^k \in \mathbb{R}^n \setminus S$ , s.t.  $y^k \to x$  as  $k \to \infty$ . We leave it as an exercise to prove that if  $x^k = P_S(y^k)$ , then  $x^k \to x$ ,  $k \to \infty$ .

Let  $g^k = \frac{y^k - x^k}{\|y^k - x^k\|}$ . Then, by part 5, we have  $g^k \in N_S(x^k)$ . Without loss of generality, we can assume that  $g^k \to g \in \mathbb{R}^n$ , with  $\|g\| = 1$ .

We claim that  $g \in N_S(x)$ . To prove the claim, note that since  $g^k \in N_S(x^k)$ , we have  $(g^k)^T(z-x^k) \leq 0$ , and

$$g^{T}(z-x) = (g-g^{k})^{T}(z-x) + (g^{k})^{T}(z-x) = (g-g^{k})^{T}(z-x) + (g^{k})^{T}(x^{k}-x) + (g^{k})^{T}(z-x^{k}) \leq (g-g^{k})^{T}(z-x) + (g^{k})^{T}(x^{k}-x) \rightarrow 0 \text{ as } k \rightarrow \infty$$

So for all  $z \in S$ , we have  $g^T(z - x) \leq 0$ , which means  $g \in N_S(x)$ . Obviously  $0 \in N_S(x)$ , so  $\{0\} \subseteq N_S(x)$ .

Proposition 5 shows that normal cones detect the boundary and interior of convex sets.

**Corollary 5** Let  $S \subseteq \mathbb{R}^n$  be a nonempty, closed, convex set. Then

- $N_S(x) = \{0\}$  if, and only if,  $x \in int(S)$ .
- $N_S(x) \supseteq \{0\}$  if, and only if,  $x \in S \setminus int(S)$ .