## 2 Review of convexity (continued)

We saw last time that any closed bounded convex set in  $\mathbb{R}^n$  is the convex hull of its extreme points (Minkowski's theorem). In the next theorem we show that any closed convex set can be expressed as an intersection of halfspaces.

**Theorem 2.1.** Assume C is a closed convex subset of  $\mathbb{R}^n$ . Then C is the intersection of all halfspaces that contain it, i.e., we have

$$C = \bigcap_{\substack{H \text{ halfspace}\\C \subseteq H}} H.$$
 (1)

*Proof.* This is a direct application of the separating hyperplane theorem. Let the right-hand side of (1) be D. It is trivial that  $C \subseteq D$ . To show that  $D \subseteq C$  we proceed by contrapositive, i.e., we will show that if  $x \notin C$  then  $x \notin D$ . If  $x \notin C$  by the separating hyperplane theorem there exists  $a \in \mathbb{R}^n \setminus \{0\}$  and  $b \in \mathbb{R}$  such that  $\langle a, z \rangle < b$  for  $z \in C$  and  $\langle a, x \rangle > b$  (we used here a *strict* version of the separating hyperplane theorem which holds when C is closed). If we call H the halfspace  $H = \{z \in \mathbb{R}^n : \langle a, z \rangle \leq b\}$  we have  $C \subseteq H$ . Thus by definition of D we have  $D \subseteq H$ . Since  $x \notin H$  it follows  $x \notin D$  which is what we wanted.

**Summary:** We have thus seen that any closed and bounded convex subset of  $\mathbb{R}^n$  has two *dual* descriptions: an "internal" description as a convex hull of points (Minkowski's theorem); and an "external" description as an intersection of halfspaces (Theorem 2.1). This is a manifestation of *duality theory* in convex analysis/geometry.

**Definition 2.1** (Cone). A set  $K \subseteq \mathbb{R}^n$  is called a *cone* if for any  $x \in K$  and  $\lambda \ge 0$  we have  $\lambda x \in K$ . The cone is called *pointed* if  $K \cap (-K) = \{0\}$ .

Note that a set K is a convex cone if and only if for any  $x, y \in K$ ,  $x + y \in K$ . A conic combination of a set of vectors  $v_1, \ldots, v_k \in \mathbb{R}^n$  is a linear combination of the form  $\lambda_1 v_1 + \cdots + \lambda_n v_n$  where  $\lambda_1, \ldots, \lambda_k$  are nonnegative. Examples of convex cones:

- Nonnegative orthant:  $\mathbb{R}^n_+ = \{x \in \mathbb{R}^n : x_i \ge 0 \ \forall i = 1, \dots, n\}.$
- "Ice-cream cone":  $\mathbf{Q}^{n+1} = \{(x,t) \in \mathbb{R}^n \times \mathbb{R} : ||x||_2 \le t\}.$

**Definition 2.2** (Dual cone). If K is a cone in  $\mathbb{R}^n$  the dual cone  $K^*$  is defined as:

$$K^* = \{ y \in \mathbb{R}^n : \langle y, x \rangle \ge 0 \ \forall x \in K \}$$

$$\tag{2}$$

**Theorem 2.2.** Let K be a cone in  $\mathbb{R}^n$ . Then  $K^*$  is a closed convex cone. Furthermore if K itself is closed and convex then  $(K^*)^* = K$ .

*Proof.* Note that  $K^*$  can be expressed as

$$K^* = \bigcap_{x \in K} \underbrace{\{y \in \mathbb{R}^n : \langle y, x \rangle \ge 0\}}_{H_x}.$$

Each  $H_x$  is a closed halfspace, thus  $K^*$  is closed and convex as an intersection of closed convex sets. The proof that  $(K^*)^* = K$  when K is closed and convex is similar to the proof of Theorem 2.1. We leave it as an exercise.

We now define *extreme rays* of cones, which play the same role as *extreme points* for bounded closed convex sets.

**Definition 2.3** (Extreme ray of a cone). An *extreme ray* of a cone  $K \subseteq \mathbb{R}^n$  is a subset  $S \subseteq K$  of the form  $S = \mathbb{R}_+ v = \{\lambda v : \lambda \ge 0\}$  where  $v \ne 0$  that satisfies the following: for any  $x, y \in K$  if  $x + y \in S$  then  $x, y \in S$ .

We will sometimes abuse notation and say that a vector  $v \in K \setminus \{0\}$  is an *extreme ray* of K if  $\mathbb{R}_+ v$  is an extreme ray of K.

**Definition 2.4** (Conical hull). Let  $S \subseteq \mathbb{R}^n$ . The *conical hull* of S is the smallest convex cone that contains S, i.e.,

$$S = \bigcap_{\substack{K \text{ convex cone}\\S \subseteq K}} K = \left\{ x \in \mathbb{R}^n : \exists k \in \mathbb{N}_{\geq 1}, s_1, \dots, s_k \in S, \lambda_1, \dots, \lambda_k \in \mathbb{R}_{\geq 0} \text{ s.t. } x = \sum_{i=1}^k \lambda_i s_i \right\}.$$

Minkowski's theorem for cones can then be stated as:

**Theorem 2.3** (Minkowski's theorem for closed convex pointed cones). Assume K is a closed and pointed convex cone in  $\mathbb{R}^n$ . Then K is the conical hull of its extreme rays, i.e., any element in K can be expressed as a conic combination of its extreme rays.

*Proof.* See Exercise 2.2 for a proof.

**Exercise 2.1** (Properties of cones and their duals). Let K be a closed convex cone in  $\mathbb{R}^n$ .

- 1. Show that the following are equivalent:
  - (i) K has nonempty interior
  - (*ii*) span(K) =  $\mathbb{R}^n$
  - (iii) For any  $w \in \mathbb{R}^n \setminus \{0\}$  there exists  $x \in K$  such that  $\langle w, x \rangle \neq 0$ .
- 2. Show that K is pointed if and only  $K^*$  has nonempty interior.
- 3. Show that  $y \in int(K^*)$  if and only if  $\langle y, x \rangle > 0$  for all  $x \in K \setminus \{0\}$ .

**Exercise 2.2** (Proof of Minkowski's theorem for closed convex pointed cones). In this exercise we are going to prove Theorem 2.3. Let K be a closed pointed convex cone.

- We are first going to assume that there exists y ∈ ℝ<sup>n</sup> such that ⟨y,x⟩ > 0 for all x ∈ K \ {0}. Show how to prove the theorem in this case. (hint: define C = {x ∈ K s.t. ⟨y,x⟩ = 1}; show that C is a compact convex set and apply Minkowski's theorem for compact convex sets to C).
- 2. Use Exercise 2.1 to show that when K is a closed pointed convex cone, such a  $y \in \mathbb{R}^n$  verifying  $\langle y, x \rangle > 0$  for all  $x \in K \setminus \{0\}$  exists.

**Exercise 2.3** (Examples of cones). For each of the following sets: show that it is a closed convex pointed cone with nonempty interior, identify the extreme rays and give a simple expression for the dual cone:

1. 
$$\mathbb{R}^{n}_{+} = \{x \in \mathbb{R}^{n} : x_{i} \ge 0 \ \forall i = 1, \dots, n\}$$

2. 
$$\mathbf{Q}^3 = \{(x_1, x_2, t) \in \mathbb{R} \times \mathbb{R} \times \mathbb{R}_+ : \sqrt{x_1^2 + x_2^2} \le t\}$$

3.  $K = \{(x, y, z) \in \mathbb{R}^2_+ \times \mathbb{R} : \sqrt{xy} \ge |z|\}$ 

Bonus question: Show that there is a linear invertible map  $A : \mathbb{R}^3 \to \mathbb{R}^3$  such that  $A(\mathbf{Q}^3) = K$ .