Geometric Combinatorics

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CHAPTER 1

Convexity and polytopes

1. Convexity basics

A linear combination of elements in a set $S \subseteq \mathbb{R}^d$ is an expression of the form

$$\sum_{i=1}^{n} t_i x_i$$

where each $x_i \in S$. Such a linear combination is called an **affine combination** if $\sum_i t_i = 1$, a **conic combination** if $t_i \geq 0$ for each i, and a **convex combination** if it is conic and affine. The set of all linear, affine, conic, and convex combinations of a set S will be denoted $\mathbb{R}S$, $\mathrm{Aff}(S)$, $\mathbb{R}_{\geq 0}(S)$, and $\mathrm{Conv}(S)$. In words, we will refer to them as the linear span, the affine hull, the conic hull, and the convex hull of S.

Let us explore the geometric significance of these concepts when $S = \{x, y\}$ consists of two distinct points. When neither x nor y is the origin, the linear hull of S is the unique plane containing x, y and the origin. The affine hull of S is the unique line in \mathbb{R}^d containing x and y and the convex hull of S is the line segment between x and y. The conic hull of S is the union of all rays from the origin through a point in the convex hull of x and y.

One says that $S \subseteq \mathbb{R}^d$ is a *linear subspace* when $S = \mathbb{R}(S)$, an *affine subspace* when S = Aff(S), a *cone* when $S = \mathbb{R}_{>0}(S)$, and *convex* when S = Conv(S).

A **V-polytope** is the convex hull of a finite set of points, i.e. a set of the form $Conv(\{x_1,\ldots,x_n\})$.



FIGURE 1.1.1. The square is a polytope. The disc is convex, but not a polytope.

A *halfspace* is a set of the form $\{x \in \mathbb{R}^d : ax \leq c\}$ where $a \in (\mathbb{R}^d)^*$ and $c \in \mathbb{R}$. An *H-polyhedron* is an intersection of finitely many halfspaces. An *H-polytope* is a bounded H-polyhedron. We will see in Section 3 that every V-polytope is an H-polytope and vice versa. We now give several examples of polytopes.

(1) **Simplices:** Fix an integer $d \geq 1$ and for each $1 \leq i \leq n$ define $e_i \in \mathbb{R}^d$ to be the i^{th} standard basis vector in \mathbb{R}^d . We define the **standard simplices** as follows

$$\Delta_{d-1} := \text{Conv}\{e_1, \dots, e_d\}$$

$$= \{x \in \mathbb{R}^d : \sum_{i=1}^d x_i = 1; 0 \le x_i \forall i = 1, \dots, d\}.$$

(2) Cubes: Given $d \ge 1$, define the **standard** d-dimensional cube by

$$C_d := \text{Conv}(\{+1, -1\}^d)$$

= $\{x \in \mathbb{R}^d : -1 \le x_i \le 1 \forall i = 1, \dots, d\}.$

(3) Cross polytopes: Given $d \ge 1$, define the *standard d-dimensional cross polytope* by

$$C_d^* := \text{Conv}\{e_1, -e_1, e_2, -e_2, \dots, e_d, -e_d\}$$

= $\{x \in \mathbb{R}^d : \sum_{i=1}^d |x_i| \le 1\}.$

Perhaps the most fundamental quantity one can associate to a geometric object is its dimension. We would like a precise way to quantify the dimension of a polytope that is easy to work with. Linear spaces are just about the only thing in mathematics that have an obvious definition of their dimension, which is the size of a basis. The following lemma says that each affine space is uniquely associated to a linear space. We will then define the dimension of an affine space to be the dimension of the associated linear space. Then, with this at our disposal, we will define the dimension of a polytope to be the dimension of its affine hull. Given two subsets $S_1, S_2 \subseteq \mathbb{R}^d$, the (*Minkowski*) sum of S_1 and S_2 , denoted $S_1 + S_2$, is defined to be

$$S_1 + S_2 := \{x + y : x \in S_1, y \in S_2\}.$$

Lemma 1.1: Let $A \subseteq \mathbb{R}^d$ be an affine subspace. Then there exists a unique linear subspace $L \subseteq \mathbb{R}^d$ such that $A = L + \{b\}$ where b is an arbitrary element of A.

PROOF. Let $b \in A$ and define $L := A + \{-b\}$. We must show that L is indeed a linear subspace and that it does not depend on our choice of b. Indeed, let $x_1, x_2 \in A$ so that $x_1 - b$ and $x_2 - b$ are arbitrary elements of L. Their sum is $(x_1 + x_2 - b) - b$, which is also an element of L as $x_1 + x_2 - b$ is an affine combination of elements in A and is therefore in A itself. Now, let $x \in A$ and $t \in R$. Then t(x - b) = tx + (1 - t)b - b which is in L since $tx + (1 - t)b \in A$. Our choice of L does not depend on b since $x - c \in L$ for any $c \in A$ since x - c = (x - c + b) - b and $x - c + b \in A$.

The **dimension** of an affine subspace $A \subseteq \mathbb{R}^d$ is the size of a basis of the linear space $\{x - b \in \mathbb{R}^d : x \in A\}$ where $b \in A$. We denote this by $\dim(A)$. The dimension of a convex set $C \subseteq \mathbb{R}^d$ is $\dim(\operatorname{Aff}(C))$. Two convex sets $P \in \mathbb{R}^d$ and $Q \in \mathbb{R}^e$ are **affinely isomorphic** if there exists an affine function $f : \mathbb{R}^d \to \mathbb{R}^e$ that is a bijection between P and Q. The polytopes in Figure 1.1.2 suggest that the notion of affine isomorphism is too strong for combinatorics since all three polytopes are, in a sense that we will make precise soon, "combinatorially equivalent" in the sense that they both have four edges and four vertices.

2. The relative boundary of a convex set

The interesting combinatorics of a convex set happens on its relative boundary, a topological notion we will recall soon. In particular, the relative boundary of a convex set is made up of lower-dimensional convex sets, called *faces*, that form a partially ordered set under inclusion. Since we are focusing on relative boundaries, we will often restrict our consideration to closed convex sets.



FIGURE 1.1.2. The first two polytopes are affinely isomorphic to each other, but not to the third. This is because affine functions preserve parallel lines and the first two have two sets of parallel lines, but the last one has none.

We now recall the relevant topological notions. Let $S \subseteq \mathbb{R}^d$ be a set. We say that S is **closed** if it is closed under taking convergent sequences, i.e. if $x_1, x_2, \dots \in S$ and $x_n \to x$, then $x \in S$. We say that S is **open** if $\mathbb{R}^d \setminus S$ is closed, or equivalently, if S is a union of open balls. The **interior** of S is the union of all open sets contained in S and the **closure** of S is the intersection of all closed sets containing S. The **boundary** of S is the relative complement of the interior of S in the closure of S. The **relative interior** (resp. closure, boundary) of a convex set $C \subseteq \mathbb{R}^d$ is the interior (resp. closure, boundary) of S in the induced topology on AffS.

Definition 1.2: Let $C \subseteq \mathbb{R}^d$ be closed and convex. A subset $F \subseteq C$ is a **face** of C if

- (1) F is closed,
- (2) F is convex, and
- (3) given $x, y \in C$, if $ri(Conv(x, y)) \cap F \neq \emptyset$, then $x, y \in F$.

A face is called an *extreme point* if it has dimension 0, an *edge* if it has dimension 1 (or sometimes, in the case of cones, an extreme ray), and a *facet* if it has dimension $\dim(C) - 1$. A face is *proper* if it is neither C nor \emptyset .

Example 1.3: Every boundary point on a ball in \mathbb{R}^d is an extreme point and these are the only proper faces. The set of proper faces of a polygon consists of its edges and vertices. An affine space has no proper faces. The only proper face of the halfspace $\{x \in \mathbb{R}^d : ax \geq c\}$ is its boundary hyperplane, namely $\{x \in \mathbb{R}^d : ax = c\}$.

Lemma 1.4: Let $C \subseteq \mathbb{R}^d$ be convex of dimension at least 1. Then ri(C) is nonempty.

PROOF. Let k denote the dimension of C and let x_1, \ldots, x_{k+1} affinely span Aff(C). Consider the function $f: \Delta_k \to C$ given by

$$\sum_{i=1}^{k+1} t_i e_i \mapsto \sum_{i=1}^{k+1} t_i x_i.$$

Then f is continuous and injective. Since Δ_k is compact, $f(\Delta_k)$ is homeomorphic to Δ_k [1, Theorem 26.6]. Since $\frac{1}{k} \sum_{i=1}^{k+1} e_i \in \operatorname{ri}(\Delta_k)$, $f(\Delta_k)$, and therefore C, has nonempty relative interior.

Proposition 1.5: If $C \subseteq \mathbb{R}^d$ is closed and convex and $F \subset C$ is a proper face, then $F \subseteq \mathrm{rb}(C)$.

PROOF. Let $y \in F$ and let $x \in C \setminus F$. For $n = 1, 2, \ldots$, define

$$S_n := \{ ty + (1-t)x : 0 \le t \le 1 + 1/n \}.$$

Since F is a face of C and S_n is a line segment whose interior intersects F, there exists a point $y_n \in S_n \setminus C$. Then, $y_n \to y$ as $n \to \infty$. But this implies $y \in \text{rb}(C)$ because $y_n \in \text{Aff}(C)$ as $S_n \subseteq \text{Aff}(C)$.

Proposition 1.6: Let $C \subseteq \mathbb{R}^d$ be closed and convex. If F is a proper face of C, then $\dim(F) < \dim(C)$.

PROOF. Since $F \subseteq C$, $\dim(F) \leq \dim(C)$. Assume for the sake of contradiction that $\dim(F) = \dim(C)$. Since $\operatorname{Aff}(F) \subseteq \operatorname{Aff}(C)$, this implies that $\operatorname{Aff}(F) = \operatorname{Aff}(C)$. Passing to this affine hull if necessary, we may assume without loss of generality that $\dim(C) = d$. Since $\dim(F) = d$, Lemma 1.4 implies that $\operatorname{ri}(F)$ is a nonempty open subset of \mathbb{R}^d . Therefore $\operatorname{ri}(F) \subseteq \operatorname{ri}(C)$ implies $\operatorname{ri}(F) \subseteq \operatorname{ri}(C)$. But this contradicts Lemma 1.5.

Proposition 1.7: Let C be closed and convex and let $F \subseteq C$ be a face. Then:

- (1) every face of F is a face of C, and
- (2) every face of C contained in F is a face of C.

PROOF. Let F' be a face of F. Let $x, y \in C$ be such that $\operatorname{ri}(\operatorname{Conv}(x, y)) \cap F' \neq \emptyset$. Then $x, y \in F$ since F is a face of C. This implies $x, y \in F'$ as F' is a face of F. So F' is a face of C. Now let F' be a face of C contained in F. Let $x, y \in F$ with $F' \cap \operatorname{ri}(\operatorname{Conv}(x, y)) \neq \emptyset$. Since

F' is a face of C, this implies $x, y \in F'$. So F' is a face of C.

Lemma 1.8: Let $C \subseteq \mathbb{R}^d$ be closed and convex, let $a \in (\mathbb{R}^d)^*$, let $c \in \mathbb{R}$, and assume $ax \leq c$ for all $x \in C$. The set

$$F_{a,c} := \{x \in C : ax = c\}$$

is a face of C.

PROOF. Let $x, y \in C$ and assume that there exists $z \in F_{a,c} \cap ri(Conv(x,y))$. Let $t \in [0,1]$ such that z = tx + (1-t)y. Then

$$c = az = tax + (1-t)au$$
.

Since $ax \ge c$ and $ay \ge c$, this implies ax = ay = c, i.e. that $x, y \in F_{a,c}$.

A face F of a closed convex set $C \subseteq \mathbb{R}^d$ is called **exposed** if $F = F_{a,c}$ as in Lemma 1.8. The geometric interpretation of an exposed face is as follows. If $ax \leq c$ for all $x \in C$, then the hyperplane $\{x \in \mathbb{R}^d : ax = c\}$ lies tangent to C. The intersection of this hyperplane with C is the face $F_{a,c}$. A convex set may have faces that are not exposed - see Figure 1.2.3, for example. That said, we will eventually see that all faces of a polytope are exposed so we will not spend much time talking about non-exposed faces.



FIGURE 1.2.3. All of the faces of the above convex set $C \subset \mathbb{R}^2$ are exposed, aside from the four extreme points indicated by black dots. To see this, note that the tangent line to C at any one of these points will intersect along the entire edge that it lies on.

We now come to the first big theorem in convexity theory: the hyperplane separation theorem. There are various similar theorems that go by the same name and we will stick with the one that has the exact level of generality we need. The second homework guides you through a proof of this theorem. A proof will be added to these notes after that assignment is turned in.

Theorem 1.9 (Hyperplane separation theorem): Given a convex $C \subset \mathbb{R}^n$ and a point $y \in \mathbb{R}^d \setminus C$, there exists $a \in (\mathbb{R}^n)^*$ and $b \in \mathbb{R}$ such that $ax \leq b$ for all $x \in C$ and $ay \geq b$.

The geometric content of Theorem 1.9 is as follows: given a convex set $C \subseteq \mathbb{R}^d$ and a point $y \notin C$, there exists a hyperplane H containing C in one of its two half-spaces and y in the other. When $y \notin \text{rb}(C)$, H can be chosen so that neither C nor $\{y\}$ intersects H, and C and y lie on opposite sides of this hyperplane. When $y \in \text{rb}(C)$, H will contain y and be tangent to C. The hyperplane separation theorem allows us to close an important circle of ideas that will allow us to move away from topological considerations. In particular, we have the following theorem.

Theorem 1.10: Let $C \subseteq \mathbb{R}^d$ be closed and convex. Then $\mathrm{rb}(C)$ is the union of its proper faces.

PROOF. Proposition 1.5 implies that the union of the proper faces of C is contained in $\mathrm{rb}(C)$. It therefore suffices to let $x \in \mathrm{rb}(C)$ and find a proper face of C containing x. By restricting to $\mathrm{Aff}(C)$ if necessary, we may assume that $\dim(C) = d$. Since $\mathrm{ri}(C)$ is convex, Theorem 1.9 implies that there exists $a \in (\mathbb{R}^d)^*$ and $c \in \mathbb{R}$ such that ax = c and $ay \leq c$ for all $y \in C$. Lemma 1.8 implies that $F_{a,c} := \{y \in C : ay = c\}$ is a face of C and it is clear that this contains x. The dimension of $F_{a,c}$ is at most d-1 and since $\dim(C) = d$ and $F_{a,c} \neq \emptyset$, $F_{a,c}$ is proper. \square

The following theorem is often known as the Minkowski-Carathéodory theorem.

Theorem 1.11: Let $C \subset \mathbb{R}^d$ be a compact, convex set of dimension k. Then for each $x \in C$, there exist extreme points x_1, \ldots, x_{k+1} , not necessarily distinct, such that $x \in \text{Conv}(x_1, \ldots, x_{k+1})$. Moreover, one such x_i may be chosen arbitrarily.

PROOF. We induct on k. When k = 0, C is a single point and the theorem follows. Now assume $k \ge 1$ and let $x \in C$. If $x \in \text{rb}(C)$, then Theorem 1.10 implies that there exists a face F such that $x \in F$. Proposition 1.6 implies that $\dim(F) < k$ so we are done by induction.

Now suppose $x \in ri(C)$ and let x_{k+1} be an extreme point of C. Compactness of C implies that $Aff\{x, x_{k+1}\} \cap C = Conv\{x_{k+1}, y\}$ where $y \in rb(C)$. Theorem 1.10 implies that there exists a face F of C with $y \in F$ and Proposition 1.6 implies that dim(F) < k. Since $x \in ri(C)$, the definition of a face implies that $x_{k+1} \notin F$. The inductive hypothesis implies that $y \in Conv\{x_1, \ldots, x_k\}$ for extreme points x_1, \ldots, x_k of F. Proposition 1.7 implies that x_1, \ldots, x_k are also extreme points of C. Since $x \in Conv\{x_{k+1}, y\}$ and $y \in Conv\{x_1, \ldots, x_k\}$, we have that $x \in Conv\{x_1, \ldots, x_{k+1}\}$.

3. Duality and the main theorem of polytopes

The goal of this section is to prove the main theorem of polytopes, i.e. that H-polytopes are V-polytopes and vice-versa. We will do this by first showing that every H-polytope is a V-polytope. Once we have this, we will will introduce convex duality which will enable us to prove the other direction.

Each hyperplane $H \subseteq \mathbb{R}^d$ defines two halfspaces which we will denote H^+ and H^- . There is a choice to be made as to which halfspace is which, but when H is given explicitly as

$$H := \{x \in \mathbb{R}^d : ax = c\},\$$

we define

$$H^+ := \{ x \in \mathbb{R}^d : ax \le c \}$$
 and $\{ x \in \mathbb{R}^d : ax \ge c \}.$

Using this notation, each H-polytope can be written as

$$(1) \qquad \qquad \bigcap_{i=1}^{n} H_i^+$$

for hyperplanes $H_1, \ldots, H_n \subset \mathbb{R}^d$. The following lemma characterizes the extreme points of an H-polytope.

Lemma 1.12: Let P be an H-polytope as in (1), let $x \in P$, and define

$$I := \{ i \in \{1, \dots, n\} : x \in H_i \}.$$

Then x is an extreme point of P if and only if

$$\{x\} = \bigcap_{i \in I} H_i$$

PROOF. Let $a_1, \ldots, a_n \in (\mathbb{R}^d)^*$ and $c_1, \ldots, c_n \in \mathbb{R}$ such that $H_i^+ = \{a_i x \leq c_i\}$. Assume (2). Let $y, z \in \mathbb{R}^d$ with $y, z \neq x$ and $y \in P$ such that $x \in \text{ri}(\text{Conv}(y, z))$. By our hypothesis, there exists $i \in I$ such that $a_i y < c_i$. Since $a_i x = c_i$ and $x \in \text{ri}(\text{Conv}(y, z))$, it follows that $a_i z > c_i$ so $z \notin P$.

Now assume (2) fails and define $A := \bigcap_{i \in I} H_i$. Then $P \cap A$ is an H-polytope in A which we can write as $P \cap A = \{y \in A : a_i y \leq c_i \text{ for all } i \notin I\}$. We claim that $P \cap A$ is at least one-dimensional. Indeed, $P \cap A$ has the same dimension as A (which is at least one) since otherwise $P \cap A$ would lie in a hyperplane of A and so there would be some $i \notin I$ such that $a_i y = c_i$ for all $y \in P \cap A$. But $x \in P \cap A$, so this would imply $i \in I$, a contradiction. Since $a_i x < c_i$ for all $i \notin I$, $x \in \text{ri}(P \cap A)$. Therefore, there exist $y, z \in P \cap A$ such that $x \in \text{ri}(\text{Conv}(y, z))$. This implies that x is not an extreme point of P.

Corollary 1.13: Every H-polytope is a V-polytope.

PROOF. Let P be an H-polytope. Lemma 1.12 implies that P has finitely many extreme points x_1, \ldots, x_k . Theorem 1.11 then implies that $P = \text{Conv}\{x_1, \ldots, x_k\}$.

We now develop the theory of convex duality. This will enable us to use Corollary 1.13 in order to prove its converse.

Definition 1.14: Let $C \subseteq \mathbb{R}^d$. The *(polar) dual* C^* of C is

$$C^* := \{ a \in (\mathbb{R}^d)^* : ax \le 1 \text{ for all } x \in C \}.$$

We pause to note two things about our definition of duality. In particular, C need not be convex, and C^* lives in the dual of the vector space that contains C. Using the natural isomorphism between a vector space and its double dual, we may view C and C^{**} as subsets of the same space.

Theorem 1.15: Let $C \subseteq \mathbb{R}^d$. Then

- (1) C^* is closed and convex
- (2) If $D \subseteq \mathbb{R}^d$ and $C \subseteq D$, then $D^* \subseteq C^*$
- (3) $C \subseteq C^{**}$
- $(4) \ 0 \in C^*$

- (5) If $0 \in ri(C)$ then C^* is compact.
- (6) If $C \subseteq \mathbb{R}^d$ is convex, compact, and d-dimensional, then $C^{**} = C$.

PROOF. We will prove (6), leaving (1) through (5) as an exercise. We know from (3) that $C \subseteq C^{**}$, so it suffices to show that if $x \notin C$ then $x \notin C^{**}$. Theorem 1.9 implies that there exists $a \in (\mathbb{R}^d)^*$ and $c \in \mathbb{R}$ such that $ay \le c$ for all $y \in C$ and $ax \ge c$. Since C is closed, we can choose a, c so that ax > c. We may also assume that $c \ne 0$, since if c = 0, then a, ε satisfy the desired conditions for small $\varepsilon > 0$. Compactness of C implies that the functional a achieves a maximum α on C and since $0 \in C$, we know $\alpha \ge 0$. This implies c > 0 and therefore that $\frac{1}{c}ax > 1$. But this shows that $x \notin C^{**}$ since $\frac{1}{c}ay \le 1$ for all $y \in C$ (i.e. that $a \in C^*$).

Lemma 1.16: Let $P \subset \mathbb{R}^d$ be a d-dimensional V-polytope with $0 \in ri(P)$. Then P^* is a d-dimensional H-polytope and $0 \in ri(P^*)$.

PROOF. We already know from Theorem 1.15 that P^* is compact so it suffices to show that P^* is an intersection of finitely many half-spaces and that $0 \in \text{ri}(P^*)$. Assume $P = \text{Conv}\{v_1,\ldots,v_k\}$. If $a \in P^*$ then $av_i \leq 1$ for $i=1,\ldots,k$. Conversely, if $av_i \leq 1$ for all i and $x \in P$, then since $x = \sum_{i=1}^k t_i v_i$ with $\sum_{i=1}^k t_i = 1$, we have

$$ax = \sum_{i=1}^{k} t_i a v_i \le \sum_{i=1}^{k} t_i = 1$$

and therefore $a \in P^*$.

Now we argue that P^* is full-dimensional with 0 in its interior. The inequalities $av_i \leq 1$ are satisfied strictly for a=0 and therefore for all a in a small open neighborhood of 0. Thus 0 is in the interior of P^* . Since P^* has a nonempty (non-relative) interior, P^* is full-dimensional. \square

Theorem 1.17 (Main theorem of polytopes): Every H-polytope is a V-polytope and vice versa.

PROOF. In light of Corollary 1.13, it suffices to show that every V-polytope is an H-polytope. Indeed, let $P \subset \mathbb{R}^d$ be a V-polytope. By passing to Aff(P) and translating if necessary, we may assume that P is full-dimensional and that $0 \in ri(P)$. Now, Lemma 1.16 and Corollary 1.13 together tell us that P^* is a V-polytope. Applying Lemma 1.16 once more tells us that P^{**} is an H-polytope. Theorem 1.15 then tells us that $P = P^{**}$ so that P is an H-polytope as well. \square

Theorem 1.17 has earned its title as the main theorem of polytopes because many fundamental properties of polytopes are easy to prove using one of the two equivalent notions and hard using the other. Consider for example the following proposition, which has a very short proof in light of Theorem 1.17, but would otherwise be much harder if we were stuck with only one of V or H descriptions.

Proposition 1.18: Let $P,Q \subseteq \mathbb{R}^d$ be polytopes. Then P+Q and $P \cap Q$ are polytopes.

PROOF. Since P and Q are polytopes, we can write $P = \text{Conv}\{v_1, \ldots, v_n\}$ and $Q = \{u_1, \ldots, u_m\}$. We immediately see that $P + Q \supseteq \text{Conv}\{v_i + u_j : i = 1, \ldots, n; j = 1, \ldots m\}$. To see that this containment is not strict, let $x + y \in P + Q$. Then we have

$$x + y = \sum_{i=1}^{n} t_i v_i + \sum_{j=1}^{m} s_j v_j.$$

Since $\sum_{j=1}^{m} s_j = \sum_{i=1}^{n} t_i = 1$, we can rewrite the above as

$$\sum_{i=1}^{n} t_i \left(\sum_{j=1}^{m} s_j \right) v_i + \sum_{j=1}^{m} s_j \left(\sum_{i=1}^{n} t_i \right) u_j = \sum_{i=1}^{n} \sum_{j=1}^{m} t_i s_j (v_i + u_j)$$

thus showing equality.

Now we switch to an H-description. We can write

$$P = \bigcap_{i=1}^{n} H_i^+$$
 and $\bigcap_{j=1}^{m} G_j^+$

where H_i, G_j are hyperplanes. Then $P \cap Q$ is just

$$P \cap Q = \bigcap_{i=1}^{n} H_i^+ \cap \bigcap_{j=1}^{m} G_j^+$$

which represents $P \cap Q$ as an H-polytope.

4. Exercises

Problem 1: Show that every compact convex set has an extreme point. Give an example of a non-compact convex set with an extreme point.

Problem 2: Prove that $x \in Aff(x_1, ..., x_n)$ if and only if

$$\begin{pmatrix} 1 \\ x \end{pmatrix} \in \mathbb{R} \begin{pmatrix} 1 & 1 & \dots & 1 \\ x_1 & x_2 & \dots & x_n \end{pmatrix}$$

and that $x \in \text{Conv}(x_1, \dots, x_n)$ if and only if

$$\begin{pmatrix} 1 \\ x \end{pmatrix} \in \mathbb{R}_{\geq 0} \begin{pmatrix} 1 & 1 & \dots & 1 \\ x_1 & x_2 & \dots & x_n \end{pmatrix}$$

Problem 3: Let $X, Y \subseteq \mathbb{R}^d$ with |X| = |Y| = d+1. Assume X, Y are each affinely independent sets. Prove that Conv(X) and Conv(Y) are affinely isomorphic.

Problem 4: Let $P, Q \subset \mathbb{R}^2$ be two-dimensional polytopes (i.e. polygons). For each of the following statements, either prove that they are true, or provide a counterexample.

- (1) If P and Q have the same number of edges, then they are affinely isomorphic.
- (2) If P and Q have the same number of edges, then they are combinatorially isomorphic.
- (3) If P and Q are both triangles, then they are affinely isomorphic.
- (4) If P is a square and Q is a parallelogram, then P and Q are affinely isomorphic. Begin by convincing yourself that it makes no difference if you assume that the vertices of P are $\{0,1\}^2$ and that (0,0) is a vertex of Q.

Problem 5: Prove that $\dim(\operatorname{Conv}\{v_1,\ldots,v_n\}) = \operatorname{rank}(\hat{V}) - 1$ where

$$\hat{V} = \begin{pmatrix} 1 & 1 & \dots & 1 \\ v_1 & v_2 & \dots & v_n \end{pmatrix}$$

Problem 6: Prove the unproven parts of Theorem 1.15.

Problem 7: Prove that every face of a polytope is exposed. Then prove that every face of a polytope is a polytope.

Problem 8: Let $P,Q \subseteq \mathbb{R}^d$ be convex sets. Prove that $(P+Q)^* = P^* \cap Q^*$.

Problem 9: Prove that the standard cube is indeed dual to the standard cross-polytope, as the notation suggests.

CHAPTER 2

The face lattice

1. Preliminaries on partially ordered sets

A partially ordered set, or poset, is a pair (S, \leq) consisting of a set S and relation \leq on S satisfying the following properties:

- (1) Reflexivity: let $x \in S$. Then $x \leq x$.
- (2) Transitivity: let $x, y, z \in S$ such that $x \leq y$ and $y \leq z$. Then $x \leq z$.
- (3) Anti-symmetry: let $x, y \in S$. If $x \leq y$ and $y \leq x$, then x = y.

Given a poset (S, \leq) and $x, y \in S$, we use the notation x < y to mean $x \leq y$ and $x \neq y$. Two posets (S_1, \leq_1) and (S_2, \leq_2) are **isomorphic** if there exists a bijection $\phi: S_1 \to S_2$ such that for all $x, y \in S_1$, $x \leq_1 y$ if and only if $\phi(x) \leq_2 \phi(y)$. If x < y and $x \leq z \leq y$ implies z = x or z = y, then we say that x **covers** y and denote this by x < y. A partial order satisfying the additional proper that $x \leq y$ or $y \leq x$ for all $x, y \in S$ is called a **total order**. When S is finite, we can represent (S, \leq) using an **order diagram**. This is a directed graph G whose vertices correspond to elements of S and has an arc $x \to y$ whenever x < y. When drawing an order diagram, the convention is to draw things so that all arcs are oriented up, and then not put arrows on the edges.

Example 2.1: The subsets of $\{1, 2, ..., n\}$ form a partially ordered set when ordered by inclusion. We denote this poset by B_n and call it a **boolean lattice**. Given $S, T \subseteq \{1, ..., n\}, S \leqslant T$ if and only if $S \subseteq T$ with $|T \setminus S| = 1$. The order diagram of B_3 is shown in Figure 2.1.1.



FIGURE 2.1.1. The boolean lattice B_3 is the set of all subsets of $\{1, 2, 3\}$, partially ordered by inclusion. Its order diagram is shown in this figure. This lattice is both atomic and coatomic.

Given a poset (S, \leq) and $x, y, z \in S$, we say that z is an **upper bound** of x, y if $z \geq x$ and $z \geq y$ and a **least upper bound** if additionally $z \leq w$ for any other upper bound w of x, y.



FIGURE 2.1.2. The order diagram of a poset that is not a lattice.

Lower bounds and **greatest lower bounds** are defined analogously. Least upper bounds and greatest lower bounds are unique (see Problem 1). If every pair $x, y \in S$ have both a least upper bound and greatest lower bound, then (S, \leq) is called a **lattice**. Not every poset is a lattice - see Figure 2.1.2.

When (S, \leq) is a lattice, we denote the least upper bound of x, y by $x \vee y$ and call it their **join**, and the greatest lower bound by $x \wedge y$ and call it the **meet**. Unsurprisingly, boolean lattices are indeed lattices and the join of two elements is their union and the meet is their intersection. The meet and join operations of any lattice each satisfy an associative law and together satisfy two absorption laws. In fact, these two algebraic axioms are enough to completely characterize lattices - see Problem 4.

Associativity of the meet and join operations allows us to extend them to arbitrary sets. In particular, one can define

$$\bigvee_{i=1}^{n} x_i := x_1 \vee x_2 \vee \dots \vee x_n \quad \text{and} \quad \bigwedge_{i=1}^{n} x_i := x_1 \wedge x_2 \wedge \dots \wedge x_n$$

for arbitrary finite sets. These operations can also be extended to infinite sets, but we will not encounter any.

An element x of a poset (S, \leq) is called a **one-hat** if $x \geq y$ for all $y \in S$ and a **zero-hat** if $x \leq y$ for all $y \in S$. We denote these symbolically by $\hat{1}$ and $\hat{0}$ and every finite lattice has one of each (Problem 2). When (S, \leq) has a $\hat{0}$, elements covering $\hat{0}$ are called **atoms** and when it has a $\hat{1}$, elements covered by $\hat{1}$ are called **coatoms**. A lattice is **atomic** if every non- $\hat{0}$ element can be expressed as a join of atoms and **coatomic** if every non- $\hat{1}$ element can be expressed as a meet of coatoms. Boolean lattices B_n are both atomic and coatomic for all n. Every finite total order is a lattice, but if it has four or more elements, then it is neither atomic nor coatomic.

In the next section, we will define a partially ordered set that one can associate to any convex set, then show that in the case of polytopes, this poset is a lattice that is both atomic and coatomic. The meet operation will be relatively easy to work with, but the join operation less so. For this reason, we will need the following lemma that will enable us to assert the existence of a join operation without having to work with it explicitly.

Lemma 2.2: Let (S, \leq) be a finite poset with a $\hat{1}$ such that every pair of elements has a greatest lower bound. Then every pair of elements has a least upper bound so (S, \leq) is a lattice.

PROOF. Let $x, y \in S$ and define T to be the set of all upper bounds of x, y, i.e.

$$T := \{ z \in S : z \ge x \text{ and } z \ge y \}.$$

Since (S, \leq) contains a $\hat{1}$, T is nonempty. Since S, and therefore T, is finite and each pair of elements in T contains a greatest lower bound in S, there exist a greatest lower bound $w \in S$ of T. Since x and y are both lower bounds of T, $x \leq w$ and $y \leq w$. In other words $w \in T$, i.e. w



FIGURE 2.2.3. A lattice that is not graded.

is an upper bound of x and y. Since T is the set of all upper bounds of x and y, this implies w is the least upper bound of x and y.

2. The face lattice of a polytope

For each convex set $C \subseteq \mathbb{R}^d$, we let $\mathcal{F}(C)$ denote the set of all faces of C, partially ordered by inclusion. We call $\mathcal{F}(C)$ the **face poset of** C, and when C is a polytope, the **face lattice of** C. As we shall soon see, our use of the word "lattice" is justified. Face lattices are a fundamental concept in the study of polytopes because they allow us to rigorously define the notion of combinatorial equivalence of polytopes. In particular, two polytopes are **combinatorially equivalent** if they have isomorphic face lattices.

The main goal of this section is to prove Theorem 2.3, a structure theorem for the face lattice of a polytope. In order to do so, we must introduce a few more poset terms. Given a poset (S, \leq) and $x, y \in S$ satisfying $x \leq y$, we define the **interval between** x, y to be

$$[x,y]:=\{z\in S:x\leq z\leq y\}.$$

The *opposite* of a poset (S, \leq) , is the poset (S, \leq) where $x \leq y$ if and only if $y \leq x$. A finite lattice is *graded* if every maximal totally ordered subset has the same cardinality. In a graded lattice (P, \leq) , we define the *rank function* $r: S \to \mathbb{N}$ of (P, \leq) recursively by

$$r(x) := \begin{cases} 0 & \text{if } x = \hat{0} \\ r(y) + 1 & \text{if } y \leqslant x. \end{cases}$$

This is well-defined because (S, \leq) is graded. The **rank** of a graded lattice is the rank of $\hat{1}$. The boolean lattice B_n is graded of rank n and the rank of each $S \subseteq \{1, \ldots, n\}$ is |S|. See Figure 2.2.3 for an example of a lattice that is not graded.

Theorem 2.3: Let $P \subseteq \mathbb{R}^d$ be a k-dimensional polytope. Then:

- (1) $\mathcal{F}(P)$ is an atomic and coatomic graded lattice of rank k+1
- (2) The rank function of $\mathcal{F}(P)$ is given by $F \mapsto \dim(F) + 1$
- (3) Given $F, G \in \mathcal{F}(P)$ with $F \subseteq G$, [F, G] is the face lattice of a polytope of dimension $\dim(G) \dim(F) 1$
- (4) The opposite poset of $\mathcal{F}(P)$ is $\mathcal{F}(P^*)$.

We will break Theorem 2.3 and its proof into several smaller lemmas.

Lemma 2.4: Let $C \subseteq \mathbb{R}^d$ be convex. For any pair F_1, F_2 of faces of C, $F_1 \cap F_2$ is a face of C.

PROOF. Let F_1, F_2 be faces of C and let $F := F_1 \cap F_2$ Let $x \in F$ and let $y, z \in C$ such that $x \in \text{ri}(\text{Conv}(y, z))$. Since F_1 is a face, this implies $y, z \in F_1$ and similarly for F_2 . So $y, z \in F$ and therefore F is a face of C.

Lemma 2.5: Each polytope has finitely many faces, each of which is itself a polytope.

PROOF. Let $P \subseteq \mathbb{R}^d$ be a polytope and let $F \subseteq P$ be a face. Proposition 1.7 implies that every extreme point of F is an extreme point of P. Lemma 1.12 implies that P, and therefore F, has only finitely many extreme points. Theorem 1.11 implies that F is the convex hull of these finitely many extreme points, i.e. is a polytope. This argument also shows that the number of faces of P is bounded above by 2^k where k is the number of extreme points of P. In particular, the number of faces of P is finite.

Proposition 2.6: Let $P \subseteq \mathbb{R}^d$ be a polytope. Then $\mathcal{F}(P)$ is a finite atomic lattice. Moreover, given $F_1, F_2 \in \mathcal{F}(P), F_1 \wedge F_2 = F_1 \cap F_2$.

PROOF. That $\mathcal{F}(P)$ is a finite lattice with meet operation given by intersection is an immediate consequence of Lemmas 2.4, 2.5, and 2.2, and that $F_1 \cap F_2$ is the maximal face of P contained in both F_1 and F_2 .

We now argue that $\mathcal{F}(P)$ is atomic. Since the atoms of $\mathcal{F}(P)$ are the extreme points of P, it suffices to show that each face F of P is the minimal face containing all of its extreme points. Theorem 1.11 implies that F is the convex hull of its extreme points. Since the convex hull of a set S is the minimal convex set containing S, this implies that F is the minimal subset of P containing all of its extreme points. Since F is a face, this implies that F is moreover the minimal face of P with this property.

Since $\mathcal{F}(P)$ is an atomic lattice with finitely many atoms, $\mathcal{F}(P)$ is finite.

Lemma 2.7: Let $P \subseteq \mathbb{R}^d$ be a d-dimensional polytope with $0 \in ri(P)$ so that P^* is a polytope (c.f. Lemma 1.16). For each face F of P, define

$$F' := \{ a \in P^* : ax = 1 \text{ for all } x \in F \}.$$

Then the following hold:

- (1) F' is a face of P^* ,
- (2) the map $F \mapsto F'$ is a bijection, and
- (3) if F, G are faces of P with $F \subseteq G$, then $G' \subseteq F'$.

PROOF. Let $a, b \in (\mathbb{R}^d)^*$ so that there exists $c \in \operatorname{ri}(\operatorname{Conv}(a, b)) \cap F'$. Assume $a \in P^*$. Then $ax \leq 1$ for all $x \in P$. If $a \notin F'$, then there exists $x \in F$ such that ax < 1. But since cx = 1 and c = ta + (1 - t)b for some $0 \leq t \leq 1$, this would imply bx > 1 and therefore that $b \notin P^*$. So F' is a face of P^* .

For the second claim, note that we can apply this construction to the faces of P^* . In particular, (F')' = F and so the map $F \mapsto F'$ is injective. By the same logic, the same map applied to the faces of P^* is injective. Since P and P^* have finitely many faces by Lemma 2.5, this implies that the map $F \mapsto F'$ is a bijection.

The third claim is immediate. \Box

Corollary 2.8: Let $P \subseteq \mathbb{R}^d$ be a d-dimensional polytope with $0 \in ri(P)$. Then $\mathcal{F}(P^*)$ is the opposite poset of $\mathcal{F}(P)$ and so $\mathcal{F}(P)$ is coatomic.

3. Exercises

Problem 1: Show that least upper bounds and greatest lower bounds in a poset are unique.

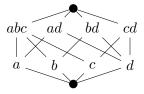
Problem 2: Prove that every finite lattice has a $\hat{0}$ and $\hat{1}$.

Problem 3: For each of the following posets, determine which are lattices. Among those that are, determine which are atomic and/or coatomic.









Problem 4: An *algebraic lattice* consists of a set S and two binary operations \vee and \wedge satisfying the following two axioms:

- (1) $x \lor (y \lor z) = (x \lor y) \lor z$ and $x \land (y \land z) = (x \land y) \land z$ for all $x, y, z \in S$ (associativity)
- (2) $x \vee (x \wedge y) = x$ and $x \wedge (x \vee y)$ for all $x, y \in S$ (absorption).

Show that if (S, \leq) is a lattice with join and meet operations \vee and \wedge , then (S, \vee, \wedge) is an algebraic lattice. Then, show that if (S, \vee, \wedge) is an algebraic lattice, then there exists a partial order \leq on S that is a lattice with meet and join operations \vee and \wedge .

Problem 5: Define a partial order \prec on \mathbb{N} by $x \prec y$ if and only if every prime divisor of x is a prime divisor of y.

- (1) Show that (\mathbb{N}, \prec) is a lattice.
- (2) Does (\mathbb{N}, \prec) have a $\hat{0}$ and/or a $\hat{1}$? If applicable, determine its atoms/coatoms.
- (3) Is (\mathbb{N}, \prec) atomic and/or coatomic?
- (4) Show that (\mathbb{N}, \prec) is isomorphic to the poset (S, \subseteq) where S is the set of all finite multisets with elements in \mathbb{N} , partially ordered by inclusion.

Bibliography

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