

by (1.4), a contradiction. Therefore, $K = \widehat{K}$. Since $y_x \in \exp K$, for $x \in \mathbb{R}^n$, we obtain

$$K = \widehat{K} \subset \text{cl conv exp } K \subset K,$$

hence $K = \text{cl conv exp } K$. \square

Corollary 1.6 (Straszewicz's Theorem) *Let $K \subset \mathbb{R}^n$ be compact and convex. Then*

$$\text{ext } K \subset \text{cl exp } K.$$

Proof By Theorems 1.23 and 1.12, we have

$$K = \text{cl conv exp } K \subset \text{cl conv cl exp } K = \text{conv cl exp } K \subset K,$$

hence

$$K = \text{conv cl exp } K.$$

By Theorem 1.21, this implies that $\text{ext } K \subset \text{cl exp } K$. \square

Exercises and Supplements for Sect. 1.5

1. Let $A \subset \mathbb{R}^n$ be closed and convex. Show that $\text{ext } A \neq \emptyset$ if and only if A does not contain any line.
2. Let $K \subset \mathbb{R}^n$ be compact and convex.
 - (a) If $n = 2$, show that $\text{ext } K$ is closed.
 - (b) If $n \geq 3$, show by an example that $\text{ext } K$ need not be closed.
- 3.* Let $A \subset \mathbb{R}^n$ be closed and convex. A subset $M \subset A$ is called an *extreme set* in A (or a *face* of A) if M is convex and if $x, y \in A$, $(x, y) \cap M \neq \emptyset$ implies that $[x, y] \subset M$. The set A and \emptyset are faces of A , all other faces are called proper. Prove the following assertions.
 - (a) Extreme sets M are closed.
 - (b) Each support set of A is extreme.
 - (c) If $M, N \subset A$ are extreme, then $M \cap N$ is extreme. (This extends to arbitrary families of extreme sets.)
 - (d) If M is extreme in A and $N \subset M$ is extreme in M , then N is extreme in A .
 - (e) If $M, N \subset A$ are extreme and $M \neq N$, then $\text{relint } M \cap \text{relint } N = \emptyset$.
 - (f) Let B be a nonempty and relatively open subset of A . Then there is a unique face F of A such that $B \subset \text{relint}(F)$.

(g) Let $\mathcal{E}(A) := \{M \subset A : M \text{ extreme}\}$. Then $A = \bigcup_{M \in \mathcal{E}(A)} \text{relint } M$ is a disjoint union.

(h) Extreme sets of A of dimension $\dim(A) - 1$ are always support sets.

4. A real (n, n) -matrix $A = ((\alpha_{ij}))$ is called *doubly stochastic* if $\alpha_{ij} \geq 0$ and

$$\sum_{k=1}^n \alpha_{kj} = \sum_{k=1}^n \alpha_{ik} = 1$$

for $i, j \in \{1, \dots, n\}$. A doubly stochastic matrix with components in $\{0, 1\}$ is called a *permutation matrix*.

Prove the following statements.

- (a) The set $K \subset \mathbb{R}^{n^2}$ of doubly stochastic matrices is compact and convex.
- (b) The extreme points of K are precisely the permutation matrices.

5. Let $P \subset \mathbb{R}^n$ be a polyhedral set, but not an affine subspace (a flat). Let

$$P = \text{aff}(P) \cap \bigcap_{i=1}^m H^-(u_i, \alpha_i)$$

$$H^-(u, \alpha) := \{x : \langle x, u \rangle \leq \alpha\}$$

with $\alpha_1, \dots, \alpha_m \in \mathbb{R}$ and $u_1, \dots, u_m \in \mathbb{R}^n \setminus \{0\}$ be a representation of P in which none of the halfspaces $H^-(u_i, \alpha_i)$ can be omitted. Put $F_i := P \cap H(u_i, \alpha_i)$. Then

- (a) $\text{relint}(P) = \bigcap_{i=1}^m \{x \in P : \langle x, u_i \rangle < \alpha_i\}$.
- (b) $\text{relbd}(P) = \bigcup_{i=1}^m F_i$.
- (c) F_1, \dots, F_m are precisely the facets of P .
- (d) Each proper face F of P is equal to the intersection of all facets of P which contain F .
- (e) The number of faces of P is finite. Each face of P is a support set and polyhedral (or empty).

6. Let $P \subset \mathbb{R}^n$ be a polyhedral set, but not an affine subspace (a flat). Prove the following statements.

- (a) If $0 \leq j \leq k$, $F^j \in \mathcal{F}_j(P)$, $F^k \in \mathcal{F}_k(P)$ and $F^j \subset F^k$, then there are $F^i \in \mathcal{F}_i(P)$ for $i = j+1, \dots, k-1$ with $F^j \subset F^{j+1} \subset \dots \subset F^{k-1} \subset F^k$.
- (b) If $0 \leq j \leq k < \dim(P)$ and $F^j \in \mathcal{F}_j(P)$, then $F^j = \bigcap \{F \in \mathcal{F}_k(P) : F^j \subset F\}$.
- (c) If $\dim(P) = n$, then each $(n-2)$ -dimensional face of P is contained in precisely two facets of P .

7. Let $\emptyset \neq P \subset \mathbb{R}^n$ be closed and convex. If the number of different support sets of P is finite, then P is a polyhedral set.

8. Let $A \subset \mathbb{R}^n$ be a closed convex set with $A \neq \text{conv}(\text{relbd}(A))$. Then A is a flat or a semi-flat.
9. Let $\emptyset \neq A \subset \mathbb{R}^n$ be closed and convex. The *recession cone* of A is defined by

$$\text{rec}(A) := \{u \in \mathbb{R}^n : A + u \subset A\}.$$

Show that the recession cone of A is a closed convex cone.

For each $x \in \mathbb{R}^n$ and $u \in \mathbb{R}^n$, let $G^+(x, u) := \{x + \lambda u : \lambda \geq 0\}$. Then, for an arbitrary $x \in A$,

$$\text{rec}(A) = \{u \in \mathbb{R}^n : G^+(x, u) \subset A\}.$$

Moreover, if $G^+(x, u) \subset A$ and $y \in A$, then also $G^+(y, u) \subset A$.

10. Each closed convex set $A \subset \mathbb{R}^n$ can be written in the form $A = \bar{A} \oplus V$, where V is a linear subspace and \bar{A} is a line-free, closed convex set which is contained in a linear subspace complementary to V .
11. Let $A \subset \mathbb{R}^n$ be closed and convex. A ray $G^+(x, u) \subset A$ is called *extreme* if it is an extreme set in A . The union of all extreme rays of A is denoted by $\text{extr}(A)$. By definition, it is clear that if $G^+(x, u)$ is an extreme ray of A , then $x \in \text{ext}(A)$. Prove the following representation result.
- If $A \subset \mathbb{R}^n$ is line-free, closed and convex, then

$$A = \text{conv}(\text{ext}(A) \cup \text{extr}(A)) = \text{conv}(\text{ext}(A)) + \text{rec}(A).$$

12. A set of the form

$$\text{pos}\{a_1, \dots, a_m\} = \left\{ \sum_{i=1}^m \lambda_i a_i : \lambda_i \geq 0 \text{ for } i = 1, \dots, m \right\}$$

with $a_1, \dots, a_m \in \mathbb{R}^n$ is called a *finitely generated convex cone*.

The convex hull of finitely many points is a polytope and hence a polyhedral set (as shown before). Show the following fact, which states that the positive hull of finitely many vectors is a polyhedral set: A finitely generated convex cone in \mathbb{R}^n is a polyhedral set, in particular, it is a closed set.

13. Let $P \subset \mathbb{R}^n$ be a polyhedral set. Then there are points $a_1, \dots, a_m \in \mathbb{R}^n$, $m \geq 1$, and vectors $b_1, \dots, b_p \in \mathbb{R}^n$ with $P = \text{conv}\{a_1, \dots, a_m\} + \text{pos}\{b_1, \dots, b_p\}$. In particular, a bounded polyhedral set is a polytope.
14. Let points $a_1, \dots, a_m \in \mathbb{R}^n$ with $m \geq 1$ and vectors $b_1, \dots, b_p \in \mathbb{R}^n$ be given. Then $P := \text{conv}\{a_1, \dots, a_m\} + \text{pos}\{b_1, \dots, b_p\}$ is a polyhedral set.
15. Hints to the literature: Convex polytopes and polyhedral sets are treated in greater detail in [2, 23, 39, 40, 49, 64, 71, 93, 94]. Combinatorial aspects of convexity are in the focus of [5, 17, 29, 44, 60, 68, 71, 75, 90]. The connection between discrete and convex geometry is the subject of [17, 19, 21, 36, 38, 60, 71, 77, 95–97]. For algorithmic aspects and combinatorial geometry, see [16, 28, 29, 36, 49, 54, 74, 75].