

Introduction to Convexity

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LSSU Math 500

1 Convex Sets

- Basic Properties of Convex Sets
- The Convex Hull
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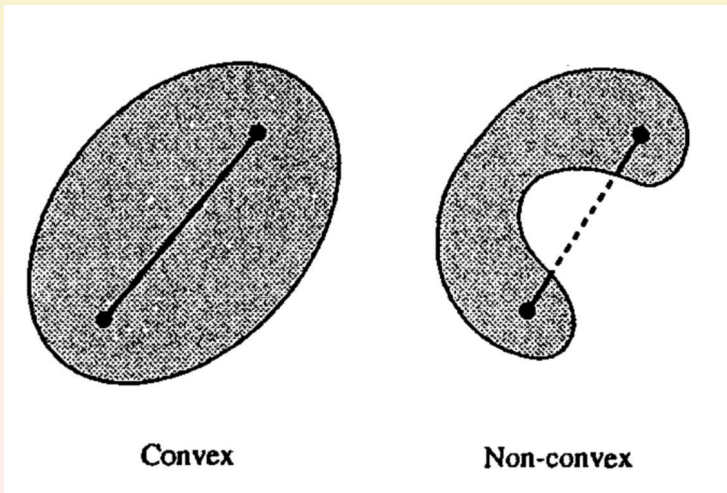
Subsection 1

Basic Properties of Convex Sets

Convex Sets in Space

- A set in space is **convex** if whenever it contains two points, it also contains the line segment joining them.
- Elementary geometry abounds in convex sets:
 - ellipses;
 - triangles;
 - parallelograms;
 - balls;
 - halfspaces;
 - cubes.
- Examples of non-convex sets are:
 - an annulus;
 - a crescent;
 - the vertex set of a cube.

Convex Sets in Space: Illustration



Convex Sets in \mathbb{R}^n

- Let \mathbf{x} and \mathbf{y} be distinct points of \mathbb{R}^n .
- Then the subset

$$\{\lambda\mathbf{x} + \mu\mathbf{y} : \lambda, \mu \geq 0, \lambda + \mu = 1\}$$

of the line through \mathbf{x} and \mathbf{y} is called the **line segment** joining \mathbf{x} and \mathbf{y} .

- The set A in \mathbb{R}^n is said to be **convex** if whenever it contains two points, it also contains the line segment joining them.
- Expressed algebraically, A is convex if $\lambda\mathbf{x} + \mu\mathbf{y} \in A$ whenever $\mathbf{x}, \mathbf{y} \in A$ and $\lambda, \mu \geq 0$ with $\lambda + \mu = 1$.
- Equivalently, A is convex if $\lambda A + \mu A \subseteq A$ whenever $\lambda, \mu \geq 0$ with $\lambda + \mu = 1$.

First Examples

- The condition for a set to be convex is less restrictive than for it to be a flat.
- So every flat is a convex set.
- In particular, the following are convex:
 - the empty set;
 - singletons;
 - lines;
 - hyperplanes;
 - \mathbb{R}^n itself.

Balls are Convex

- We show that the closed ball $B[\mathbf{a}; r]$ in \mathbb{R}^n is convex.
- Let $\mathbf{x}, \mathbf{y} \in B[\mathbf{a}; r]$.
- Let $\lambda, \mu \geq 0$ with $\lambda + \mu = 1$.
- Then $\|\mathbf{x} - \mathbf{a}\| < r$, $\|\mathbf{y} - \mathbf{a}\| < r$.

• So

$$\begin{aligned}\|\lambda\mathbf{x} + \mu\mathbf{y} - \mathbf{a}\| &= \|\lambda(\mathbf{x} - \mathbf{a}) + \mu(\mathbf{y} - \mathbf{a})\| \\ &\leq \lambda\|\mathbf{x} - \mathbf{a}\| + \mu\|\mathbf{y} - \mathbf{a}\| \\ &\leq \lambda r + \mu r = r.\end{aligned}$$

- Thus $\lambda\mathbf{x} + \mu\mathbf{y} \in B[\mathbf{a}; r]$.
- This proves that $B[\mathbf{a}; r]$ is convex.
- A similar argument shows that the open ball $B(\mathbf{a}; r)$ is convex.

Halfspaces are Convex

- We show that the closed halfspace A in \mathbb{R}^n defined by the inequality $\mathbf{u} \cdot \mathbf{x} \leq u_0$ is convex.
- Let $\mathbf{x}, \mathbf{y} \in A$ and let $\lambda, \mu \geq 0$ with $\lambda + \mu = 1$.
- Then $\mathbf{u} \cdot \mathbf{x} \leq u_0$, $\mathbf{u} \cdot \mathbf{y} \leq u_0$.

- So

$$\mathbf{u} \cdot (\lambda \mathbf{x} + \mu \mathbf{y}) = \lambda \mathbf{u} \cdot \mathbf{x} + \mu \mathbf{u} \cdot \mathbf{y} \leq \lambda u_0 + \mu u_0 = u_0.$$

- Thus $\lambda \mathbf{x} + \mu \mathbf{y} \in A$.
- This proves that A is convex.
- A similar argument shows that open halfspaces are convex.

Closure Under Intersections

Theorem

The intersection of an arbitrary family of convex sets in \mathbb{R}^n is convex.

- Let $(A_i : i \in I)$ be a family of convex sets in \mathbb{R}^n .

If $\mathbf{a}, \mathbf{b} \in \bigcap (A_i : i \in I)$ and $\lambda, \mu \geq 0$ with $\lambda + \mu = 1$, then $\mathbf{a}, \mathbf{b} \in A_i$.

As A_i is convex, $\lambda \mathbf{a} + \mu \mathbf{b} \in A_i$, for each $i \in I$.

Thus $\lambda \mathbf{a} + \mu \mathbf{b} \in \bigcap (A_i : i \in I)$.

This shows that the intersection is convex.

Closure Under Restricted Linear Combinations

Theorem

Let $\mathbf{a}_1, \dots, \mathbf{a}_m$ be points of a convex set A in \mathbb{R}^n . Let $\lambda_1, \dots, \lambda_m \geq 0$ with $\lambda_1 + \dots + \lambda_m = 1$. Then $\lambda_1 \mathbf{a}_1 + \dots + \lambda_m \mathbf{a}_m \in A$.

- We argue by induction on m .
- When $m = 1$ the assertion is trivial.
- Suppose that the assertion holds when m is some positive integer k .
- Let

$$\mathbf{x} = \lambda_1 \mathbf{a}_1 + \dots + \lambda_{k+1} \mathbf{a}_{k+1},$$

where $\mathbf{a}_1, \dots, \mathbf{a}_{k+1} \in A$ and $\lambda_1, \dots, \lambda_{k+1} \geq 0$ with $\lambda_1 + \dots + \lambda_{k+1} = 1$. At least one λ_i must be less than 1, say $\lambda_{k+1} < 1$. Write

$\mathbf{y} = \frac{\lambda_1}{\lambda} \mathbf{a}_1 + \dots + \frac{\lambda_k}{\lambda} \mathbf{a}_k$, where $\lambda = \lambda_1 + \dots + \lambda_k = 1 - \lambda_{k+1} > 0$. By the induction hypothesis, $\mathbf{y} \in A$. Since A is convex and contains both \mathbf{y} and \mathbf{a}_{k+1} , the equation $\mathbf{x} = \lambda \mathbf{y} + \lambda_{k+1} \mathbf{a}_{k+1}$ shows that $\mathbf{x} \in A$. This completes the proof by induction.

Convex Combinations

- A point \mathbf{x} is said to be a **convex combination** of points $\mathbf{a}_1, \dots, \mathbf{a}_m$ in \mathbb{R}^n if there exist scalars $\lambda_1, \dots, \lambda_m \geq 0$ with $\lambda_1 + \dots + \lambda_m = 1$ such that

$$\mathbf{x} = \lambda_1 \mathbf{a}_1 + \dots + \lambda_m \mathbf{a}_m.$$

- The preceding theorem can thus be expressed as:
Every convex combination of points of a convex set in \mathbb{R}^n belongs to that set.

Convexity, Vector Addition and Scalar Multiplication

Theorem

Let A, B be convex sets in \mathbb{R}^n and let α be a scalar. Then $A+B$ and αA are convex.

- Let $\lambda, \mu \geq 0$ with $\lambda + \mu = 1$. Since A, B are convex,

$$\begin{aligned}\lambda(A+B) + \mu(A+B) &= (\lambda A + \mu A) + (\lambda B + \mu B) \subseteq A+B; \\ \lambda(\alpha A) + \mu(\alpha A) &= \alpha(\lambda A + \mu A) \subseteq \alpha A.\end{aligned}$$

This shows that $A+B$ and αA are convex.

Corollary

Let A_1, \dots, A_m be convex sets in \mathbb{R}^n and let $\lambda_1, \dots, \lambda_m$ be scalars. Then $\lambda_1 A_1 + \dots + \lambda_m A_m$ is convex.

A Distributivity Property

Theorem

Let A be a convex set in \mathbb{R}^n and let $\lambda_1, \dots, \lambda_m \geq 0$. Then

$$(\lambda_1 + \dots + \lambda_m)A = \lambda_1 A + \dots + \lambda_m A.$$

- The result is trivial when each λ_i is zero.

Suppose that $\lambda = \lambda_1 + \dots + \lambda_m > 0$.

With the help of a previous theorem, we can deduce that

$$\begin{aligned}(\lambda_1 + \dots + \lambda_m)A &\subseteq \lambda_1 A + \dots + \lambda_m A \\ &= \lambda \left(\frac{\lambda_1}{\lambda} A + \dots + \frac{\lambda_m}{\lambda} A \right) \\ &\subseteq \lambda A \\ &= (\lambda_1 + \dots + \lambda_m)A.\end{aligned}$$

Thus $(\lambda_1 + \dots + \lambda_m)A = \lambda_1 A + \dots + \lambda_m A$.

A Cancellation Property

Theorem

Let A, B, C be sets in \mathbb{R}^n . Suppose that A is non-empty and bounded, that C is closed and convex, and that $A+B \subseteq A+C$. Then $B \subseteq C$.

- Let $\mathbf{a}_0 \in A$. If $\mathbf{b} \in B$, then $\mathbf{a}_0 + \mathbf{b} \in A+B \subseteq A+C$. So there exist $\mathbf{a}_1 \in A$, $\mathbf{c}_1 \in C$, such that $\mathbf{a}_0 + \mathbf{b} = \mathbf{a}_1 + \mathbf{c}_1$. Similarly, there exist $\mathbf{a}_2, \dots, \mathbf{a}_i \in A$ and $\mathbf{c}_2, \dots, \mathbf{c}_i \in C$ with $\mathbf{a}_1 + \mathbf{b} = \mathbf{a}_2 + \mathbf{c}_2$, ..., $\mathbf{a}_{i-1} + \mathbf{b} = \mathbf{a}_i + \mathbf{c}_i$. We add the i equations above together to deduce that

$$\mathbf{a}_0 + i\mathbf{b} = \mathbf{a}_i + \mathbf{c}_1 + \dots + \mathbf{c}_i.$$

Since C is convex, the point $\mathbf{x}_i = \frac{1}{i}(\mathbf{c}_1 + \dots + \mathbf{c}_i)$ lies in C . Since A is bounded,

$$\begin{aligned} \|\mathbf{b} - \mathbf{x}_i\| &= \left\| \frac{1}{i}(\mathbf{a}_i + \mathbf{c}_1 + \dots + \mathbf{c}_i - \mathbf{a}_0) - \frac{1}{i}(\mathbf{c}_1 + \dots + \mathbf{c}_i) \right\| \\ &= \frac{1}{i} \|\mathbf{a}_i - \mathbf{a}_0\| \rightarrow 0 \text{ as } i \rightarrow \infty. \end{aligned}$$

Thus $\mathbf{x}_i \rightarrow \mathbf{b}$ as $i \rightarrow \infty$. But C is closed. So $\mathbf{b} \in C$. Hence, $B \subseteq C$.

A Cancellation Property (Cont'd)

Corollary

Let A, B, C be sets in \mathbb{R}^n . Suppose that A is non-empty and bounded, that B and C are closed and convex, and that $A + B = A + C$. Then $B = C$.

Affine Transformations and Convex Sets

Theorem

Let $f: \mathbb{R}^n \rightarrow \mathbb{R}^m$ be an affine transformation. Then $f(A)$ is convex for each convex set A in \mathbb{R}^n , and $f^{-1}(B)$ is convex for each convex set B in \mathbb{R}^m .

- Let A be a convex set in \mathbb{R}^n . Let $\lambda, \mu \geq 0$ with $\lambda + \mu = 1$. If $\mathbf{x}, \mathbf{y} \in f(A)$, then $\mathbf{x} = f(\mathbf{a})$, $\mathbf{y} = f(\mathbf{b})$ for some $\mathbf{a}, \mathbf{b} \in A$. Since A is convex, $\lambda\mathbf{a} + \mu\mathbf{b} \in A$. Since f is affine,

$$\lambda\mathbf{x} + \mu\mathbf{y} = \lambda f(\mathbf{a}) + \mu f(\mathbf{b}) = f(\lambda\mathbf{a} + \mu\mathbf{b}).$$

Thus $\lambda\mathbf{x} + \mu\mathbf{y} \in f(A)$. This shows that $f(A)$ is convex.

Let B be a convex set in \mathbb{R}^m . Let $\lambda, \mu \geq 0$ with $\lambda + \mu = 1$. If $\mathbf{x}, \mathbf{y} \in f^{-1}(B)$, then $f(\mathbf{x}), f(\mathbf{y}) \in B$. Since B is convex, $\lambda f(\mathbf{x}) + \mu f(\mathbf{y}) \in B$. Since f is affine, $f(\lambda\mathbf{x} + \mu\mathbf{y}) = \lambda f(\mathbf{x}) + \mu f(\mathbf{y}) \in B$. Thus $\lambda\mathbf{x} + \mu\mathbf{y} \in f^{-1}(B)$. This shows that $f^{-1}(B)$ is convex.

Subsection 2

The Convex Hull

The Convex Hull

- The **convex hull** $\text{conv}A$ of a set A in \mathbb{R}^n is the intersection of all convex sets in \mathbb{R}^n containing A .
- The definition of $\text{conv}A$, together with a previous theorem, shows that $\text{conv}A$ is a convex set containing A .
- Moreover, if C is any convex set in \mathbb{R}^n containing A , then $\text{conv}A \subseteq C$.
- Thus we may refer to $\text{conv}A$ as the smallest convex set in \mathbb{R}^n containing A .
- Clearly, A is convex if and only if $A = \text{conv}A$.
- Moreover $\text{conv}(\text{conv}A) = \text{conv}A$.
- Also $\text{conv}A \subseteq \text{conv}B$ whenever $A \subseteq B$.

Examples

- In space:
 - The convex hull of two distinct points is the line segment joining them;
 - The convex hull of three non-collinear points is the triangle which they determine;
 - The convex hull of four non-coplanar points is the tetrahedron which they determine.
- In \mathbb{R}^2 the convex hull of m points symmetrically placed on the circumference of a circle, where $m \geq 3$, is a regular m -sided polygon.

Example

- The convex hull of the set $A = \{\mathbf{x} \in \mathbb{R}^n : \|\mathbf{x}\| = 1\}$ is the closed unit ball $U = \{\mathbf{x} \in \mathbb{R}^n : \|\mathbf{x}\| \leq 1\}$.
- The ball U is convex and contains A , so $\text{conv}A \subseteq U$.
- We now show that $U \subseteq \text{conv}A$.

Let $\mathbf{x} \in U$. If $\mathbf{x} = 0$ and $\mathbf{y} \in A$, then $\mathbf{x} = \frac{1}{2}\mathbf{y} + \frac{1}{2}(-\mathbf{y})$. Since $\text{conv}A$ is convex and contains \mathbf{y} and $-\mathbf{y}$, this shows that $\mathbf{x} \in \text{conv}A$. If $\mathbf{x} \neq 0$, then $0 < \|\mathbf{x}\| \leq 1$. The equation

$$\mathbf{x} = \left(\frac{1 + \|\mathbf{x}\|}{2}\right) \frac{\mathbf{x}}{\|\mathbf{x}\|} + \left(\frac{1 - \|\mathbf{x}\|}{2}\right) \frac{-\mathbf{x}}{\|\mathbf{x}\|}$$

shows that $\mathbf{x} \in \text{conv}A$, since $\text{conv}A$ is convex and contains $\frac{\mathbf{x}}{\|\mathbf{x}\|}$ and $-\frac{\mathbf{x}}{\|\mathbf{x}\|}$. Thus $U \subseteq \text{conv}A$.

- We now have $U = \text{conv}A$.

Description of Convex Hull in Terms of Points

Theorem

Let A be a set in \mathbb{R}^n . Then $\text{conv}A$ is the set of all convex combinations of points of A .

- Denote by B the set of all convex combinations of points of A . That $B \subseteq \text{conv}A$ follows from a previous theorem and the inclusion $A \subseteq \text{conv}A$.

We next show that B is convex. If $\mathbf{x}, \mathbf{y} \in B$, then

$$\mathbf{x} = \lambda_1 \mathbf{a}_1 + \cdots + \lambda_m \mathbf{a}_m, \quad \mathbf{y} = \mu_1 \mathbf{b}_1 + \cdots + \mu_p \mathbf{b}_p,$$

for some $\mathbf{a}_1, \dots, \mathbf{a}_m, \mathbf{b}_1, \dots, \mathbf{b}_p \in A$ and $\lambda_1, \dots, \lambda_m, \mu_1, \dots, \mu_p \geq 0$ with $\lambda_1 + \cdots + \lambda_m = 1$ and $\mu_1 + \cdots + \mu_p = 1$.

Description of Convex Hull in Terms of Points (Cont'd)

- Let $\lambda, \mu \geq 0$ with $\lambda + \mu = 1$. Then

$$\lambda \mathbf{x} + \mu \mathbf{y} = \lambda \lambda_1 \mathbf{a}_1 + \cdots + \lambda \lambda_m \mathbf{a}_m + \mu \mu_1 \mathbf{b}_1 + \cdots + \mu \mu_p \mathbf{b}_p$$

and

$$\begin{aligned} \lambda \lambda_1 + \cdots + \lambda \lambda_m + \mu \mu_1 + \cdots + \mu \mu_p \\ &= \lambda(\lambda_1 + \cdots + \lambda_m) + \mu(\mu_1 + \cdots + \mu_p) \\ &= \lambda + \mu = 1. \end{aligned}$$

Thus $\lambda \mathbf{x} + \mu \mathbf{y} \in B$, so B is convex. Since B is convex and $B \supseteq A$, it follows that $B \supseteq \text{conv}A$. Hence $B = \text{conv}A$.

Corollary

Let $\mathbf{a}_1, \dots, \mathbf{a}_m \in \mathbb{R}^n$. Then

$$\text{conv}\{\mathbf{a}_1, \dots, \mathbf{a}_m\} = \{\lambda_1 \mathbf{a}_1 + \cdots + \lambda_m \mathbf{a}_m : \lambda_1, \dots, \lambda_m \geq 0, \lambda_1 + \cdots + \lambda_m = 1\}.$$

On the Number of Points

- The preceding theorem shows that each point of the convex hull of a set in \mathbb{R}^n is a convex combination of points of that set.
- The theorem makes no reference to the number of points in the combination.
- Carathéodory's Theorem, which is proved next, states that each point of the convex hull of an r -dimensional set can be expressed as a convex combination of $r + 1$ or fewer points of the set.
- Thus a point in the convex hull of a set in \mathbb{R}^3 is either a point of the set or belongs to a line segment, a triangle, or a tetrahedron with vertices in the set.

Carathéodory's Theorem

Theorem (Carathéodory's Theorem)

Let $\mathbf{a} \in \text{conv}A$, where A is an r -dimensional set in \mathbb{R}^n . Then \mathbf{a} can be expressed as a convex combination of $r + 1$ or fewer points of A .

- The preceding theorem shows the existence of points $\mathbf{a}_1, \dots, \mathbf{a}_m$ of A and scalars $\lambda_1, \dots, \lambda_m \geq 0$ with $\lambda_1 + \dots + \lambda_m = 1$ such that

$$\mathbf{a} = \lambda_1 \mathbf{a}_1 + \dots + \lambda_m \mathbf{a}_m.$$

We assume that this representation of \mathbf{a} is so chosen that \mathbf{a} cannot be expressed as a convex combination of fewer than m points of A .

It follows that no two of the points $\mathbf{a}_1, \dots, \mathbf{a}_m$ are equal and that $\lambda_1, \dots, \lambda_m > 0$. We prove the theorem by showing that $m \leq r + 1$.

We use a contradiction argument.

Carathéodory's Theorem (Cont'd)

- Suppose that $m > r + 1$. Then, since A is r -dimensional, the set $\{\mathbf{a}_1, \dots, \mathbf{a}_m\}$ must be affinely dependent. So there exist scalars μ_1, \dots, μ_m , not all zero, such that

$$\mathbf{0} = \mu_1 \mathbf{a}_1 + \dots + \mu_m \mathbf{a}_m, \quad \mu_1 + \dots + \mu_m = 0.$$

Let $t > 0$ be such that the scalars $\lambda_1 + \mu_1 t, \dots, \lambda_m + \mu_m t$ are nonnegative with at least one of them zero. Such a t exists since the λ 's are all positive and at least one of the μ 's is negative. The equation

$$\mathbf{a} = (\lambda_1 + \mu_1 t) \mathbf{a}_1 + \dots + (\lambda_m + \mu_m t) \mathbf{a}_m,$$

when its terms with zero coefficients are omitted, exhibits \mathbf{a} as a convex combination of fewer than m points of A . This contradiction to the minimality of m shows that $m \leq r + 1$.

Radon's Theorem

Theorem (Radon's Theorem)

Let $\mathbf{a}_1, \dots, \mathbf{a}_m \in \mathbb{R}^n$ ($m \geq n+2$). Then the set $\{1, \dots, m\}$ can be partitioned into two subsets I and J such that $\text{conv}\{\mathbf{a}_i : i \in I\}$ meets $\text{conv}\{\mathbf{a}_j : j \in J\}$.

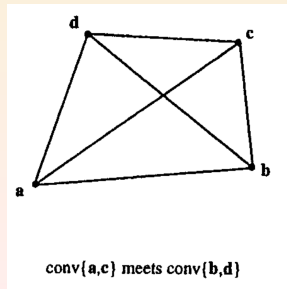
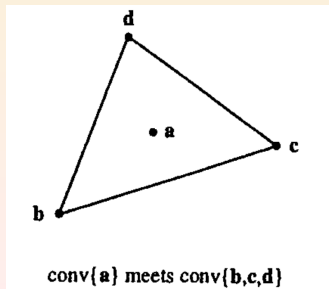
- We consider the non-trivial case when the $\mathbf{a}_1, \dots, \mathbf{a}_m$ are distinct. It follows from a previous corollary that there exist scalars $\lambda_1, \dots, \lambda_m$, not all zero, such that $\lambda_1 \mathbf{a}_1 + \dots + \lambda_m \mathbf{a}_m = \mathbf{0}$ and $\lambda_1 + \dots + \lambda_m = 0$. Some of the λ 's will be positive, others negative. Let $I = \{i : \lambda_i \geq 0\}$ and $J = \{j : \lambda_j < 0\}$. Then

$$\frac{\sum_{i \in I} \lambda_i \mathbf{a}_i}{\sum_{i \in I} \lambda_i} = \frac{\sum_{j \in J} (-\lambda_j) \mathbf{a}_j}{\sum_{j \in J} (-\lambda_j)} = \mathbf{x} \text{ say.}$$

Thus \mathbf{x} is a convex combination of points of both $\{\mathbf{a}_i : i \in I\}$ and $\{\mathbf{a}_j : j \in J\}$. Hence $\mathbf{x} \in \text{conv}\{\mathbf{a}_i : i \in I\} \cap \text{conv}\{\mathbf{a}_j : j \in J\}$.

Four Points in \mathbb{R}^2

- Radon's theorem yields information about the possible configurations of four points in \mathbb{R}^2 .
- It shows that:
 - Either one of the four points belongs to the (possibly degenerate) triangle determined by the remaining three;
 - Or the four points are the vertices of a convex quadrilateral.



Convex Hull of Open or Compact Sets

Theorem

In \mathbb{R}^n the convex hull of an open set is open and the convex hull of a compact set is compact.

- Let A be an open set in \mathbb{R}^n . If $\mathbf{a} \in \text{conv}A$, then $\mathbf{a} = \lambda_1 \mathbf{a}_1 + \cdots + \lambda_m \mathbf{a}_m$ for some $\mathbf{a}_1, \dots, \mathbf{a}_m \in A$ and $\lambda_1, \dots, \lambda_m \geq 0$ with $\lambda_1 + \cdots + \lambda_m = 1$. Since A is open, there exist $r_1, \dots, r_m > 0$ such that $B(\mathbf{a}_1; r_1) \subseteq A, \dots, B(\mathbf{a}_m; r_m) \subseteq A$. Let $r = \min\{r_1, \dots, r_m\}$, so $r > 0$. We show that $B(\mathbf{a}; r) \subseteq \text{conv}A$. Let $\mathbf{x} \in B(\mathbf{a}; r)$. Then $\|\mathbf{x} - \mathbf{a}\| < r$. For $i = 1, \dots, m$, the point $\mathbf{x}_i = \mathbf{a}_i + \mathbf{x} - \mathbf{a}$ lies in $B(\mathbf{a}_i; r)$. Hence also in $B(\mathbf{a}_i; r_i)$ and A . Now we get

$$\begin{aligned} \mathbf{x} &= \mathbf{a} + \mathbf{x} - \mathbf{a} = \lambda_1 \mathbf{a}_1 + \cdots + \lambda_m \mathbf{a}_m + (\lambda_1 + \cdots + \lambda_m)(\mathbf{x} - \mathbf{a}) \\ &= \lambda_1(\mathbf{a}_1 + \mathbf{x} - \mathbf{a}) + \cdots + \lambda_m(\mathbf{a}_m + \mathbf{x} - \mathbf{a}) = \lambda_1 \mathbf{x}_1 + \cdots + \lambda_m \mathbf{x}_m. \end{aligned}$$

So $\mathbf{x} \in \text{conv}A$. Thus $B(\mathbf{a}; r) \subseteq \text{conv}A$ and each point of $\text{conv}A$ is an interior point of $\text{conv}A$. I.e., $\text{conv}A$ is open.

Convex Hull of Open or Compact Sets (Cont'd)

- Now let A be a compact set in \mathbb{R}^n . If $\mathbf{x}_1, \dots, \mathbf{x}_k, \dots$ is a sequence in $\text{conv}A$, then, by Carathéodory's Theorem, \mathbf{x}_k can be expressed in the form $\mathbf{x}_k = \lambda_{k0}\mathbf{a}_{k0} + \dots + \lambda_{kn}\mathbf{a}_{kn}$, for some $\mathbf{a}_{k0}, \dots, \mathbf{a}_{kn} \in A$ and $\lambda_{k0}, \dots, \lambda_{kn} \geq 0$ with $\lambda_{k0} + \dots + \lambda_{kn} = 1$. It may be necessary to include some extra \mathbf{a} 's with zero coefficients to bring the number of \mathbf{a} 's in the expression for \mathbf{x}_k up to $n+1$. Each sequence $\mathbf{a}_{1j}, \dots, \mathbf{a}_{kj}, \dots$ ($j = 0, \dots, n$) belongs to the compact set A . Each real sequence $\lambda_{1j}, \dots, \lambda_{kj}, \dots$ ($j = 0, \dots, n$) belongs to the compact interval $[0, 1]$. Since there is only a finite number, namely $2n+2$, of these sequences, we can, by repeatedly forming convergent subsequences of sequences whose members lie in a compact set, find a subsequence i_1, \dots, i_k, \dots of $1, \dots, k, \dots$, points $\mathbf{a}_0, \dots, \mathbf{a}_n$ of A and scalars $\lambda_0, \dots, \lambda_n \geq 0$ with $\lambda_0 + \dots + \lambda_n = 1$, such that $\mathbf{a}_{kj} \rightarrow \mathbf{a}_j$ and $\lambda_{kj} \rightarrow \lambda_j$ ($k \rightarrow \infty, j = 0, \dots, n$). Thus the subsequence $\mathbf{x}_{i_1}, \dots, \mathbf{x}_{i_k}, \dots$ converges to the point $\lambda_0\mathbf{a}_0 + \dots + \lambda_n\mathbf{a}_n$ of $\text{conv}A$. This shows that $\text{conv}A$ is compact.

Finite Sets and Closed Sets

Corollary

The convex hull of a finite set in \mathbb{R}^n is compact.

- The theorem makes no reference to the convex hull of a closed set.
- Except in \mathbb{R}^1 , the convex hull of a closed set need not be closed.

Example: In \mathbb{R}^n the union of a line and a point not on it is a closed set.

But its convex hull is not closed.

Diameter of Bounded Sets

- Since a set in \mathbb{R}^n is bounded if and only if it lies in some ball, and balls are convex, it follows that the convex hull of a bounded set in \mathbb{R}^n is bounded.
- The **diameter** of a nonempty bounded set A in \mathbb{R}^n is the nonnegative real number

$$\sup \{ \| \mathbf{a} - \mathbf{b} \| : \mathbf{a}, \mathbf{b} \in A \}.$$

- In \mathbb{R}^2 the diameter of a triangle is the length of a longest side.
The diameter of a rectangle is the length of a diagonal.
- In \mathbb{R}^n the balls $B(\mathbf{a}; r)$ and $B[\mathbf{a}; r]$ both have diameter $2r$.
- The theorem below relates the diameters of a bounded set and its convex hull.

Diameter of a Bounded Set and its Convex Hull

Theorem

Let A be a nonempty bounded set in \mathbb{R}^n . Then A and $\text{conv}A$ have the same diameter.

- Suppose that A has diameter s . Let $\mathbf{x}, \mathbf{y} \in \text{conv}A$. Then

$$\mathbf{x} = \lambda_1 \mathbf{a}_1 + \cdots + \lambda_m \mathbf{a}_m, \quad \mathbf{y} = \mu_1 \mathbf{b}_1 + \cdots + \mu_p \mathbf{b}_p,$$

for some $\mathbf{a}_1, \dots, \mathbf{a}_m, \mathbf{b}_1, \dots, \mathbf{b}_p \in A$ and $\lambda_1, \dots, \lambda_m, \mu_1, \dots, \mu_p \geq 0$ with $\lambda_1 + \cdots + \lambda_m = 1$, $\mu_1 + \cdots + \mu_p = 1$. Thus $\mathbf{x} = \sum_{i=1}^m \sum_{j=1}^p \lambda_i \mu_j \mathbf{a}_i$ and $\mathbf{y} = \sum_{i=1}^m \sum_{j=1}^p \lambda_i \mu_j \mathbf{b}_j$. Hence, using the Triangle Inequality,

$$\begin{aligned} \|\mathbf{x} - \mathbf{y}\| &= \left\| \sum_{i=1}^m \sum_{j=1}^p \lambda_i \mu_j (\mathbf{a}_i - \mathbf{b}_j) \right\| \\ &\leq \sum_{i=1}^m \sum_{j=1}^p \lambda_i \mu_j \|\mathbf{a}_i - \mathbf{b}_j\| \\ &\leq \sum_{i=1}^m \sum_{j=1}^p \lambda_i \mu_j s = s. \end{aligned}$$

Hence the diameter of $\text{conv}A$ does not exceed s . Since $A \subseteq \text{conv}A$, the diameter of $\text{conv}A$ is at least s . Thus $\text{conv}A$ has diameter s .

Convex Hull of Set of Complex Numbers

- We now prove a result concerning the location of the roots of the derivative of a complex polynomial.
- In a natural way we can identify the Euclidean space \mathbb{R}^2 with the complex plane by identifying each point (x, y) of \mathbb{R}^2 with the complex number $x + iy$ and vice versa.
- This identification allows us to refer to the convex hull of a set of complex numbers or to a convex combination of complex numbers.

Example: Consider a complex polynomial $P(z) = az^2 + bz + c$.

Then P has roots $\frac{-b \pm \sqrt{b^2 - 4ac}}{2a}$, and its derivative P' has root $-\frac{b}{2a}$.

Hence the root of P' lies midway between the roots of P .

So the root of P' is in the convex hull of the roots of P .

The Gauss-Lucas Theorem

Theorem (Gauss-Lucas Theorem)

The roots of the derivative of a non-constant complex polynomial belong to the convex hull of the set of roots of the polynomial itself.

- Let P be the complex polynomial defined for complex z by the equation

$$P(z) = a_n z^n + \cdots + a_1 z + a_0,$$

where $n \geq 1$ and a_0, a_1, \dots, a_n are complex numbers with $a_n \neq 0$. Then

$$P(z) = a_n (z - z_1) \cdots (z - z_n),$$

where z_1, \dots, z_n are the roots of P , each being repeated according to its multiplicity. A routine verification shows that, for $z \neq z_1, \dots, z_n$,

$$\frac{P'(z)}{P(z)} = \frac{1}{z - z_1} + \cdots + \frac{1}{z - z_n} = \frac{\bar{z} - \bar{z}_1}{|z - z_1|^2} + \cdots + \frac{\bar{z} - \bar{z}_n}{|z - z_n|^2}.$$

The Gauss-Lucas Theorem (Cont'd)

- Suppose now that z is a root of P' . We establish the theorem by exhibiting z as a convex combination of z_1, \dots, z_n . This can be done trivially if z is one of z_1, \dots, z_n . So assume that this is not the case. Putting $P'(z) = 0$ in the preceding equation, we find easily that

$$z = \frac{\frac{1}{|z-z_1|^2} z_1 + \dots + \frac{1}{|z-z_n|^2} z_n}{\frac{1}{|z-z_1|^2} + \dots + \frac{1}{|z-z_n|^2}}.$$

This expresses z as a convex combination of z_1, \dots, z_n .

Corollary

Suppose that the roots of a non-constant complex polynomial lie in some given convex set. Then the roots of its derivative lie in the same convex set.

- A simple application of the corollary:
If all the roots of a non-constant complex polynomial have positive imaginary parts, then the same is also true of the roots of its derivative.

Subsection 3

Interiors and Closures

Relative Interior of a Nonempty Convex Set

Theorem

The relative interior of a non-empty convex set in \mathbb{R}^n is non-empty.

- Let A be a non-empty r -dimensional convex set in \mathbb{R}^n . Then A contains points $\mathbf{a}_0, \dots, \mathbf{a}_r$ which form an affine basis for the r -flat $\text{aff}A$, and the barycentric coordinates $\lambda_0, \dots, \lambda_r$ of a point \mathbf{x} of $\text{aff}A$ relative to $\mathbf{a}_0, \dots, \mathbf{a}_r$ are continuous functions of \mathbf{x} , a fact which follows easily from a previous theorem. Let $\mathbf{a} = \frac{1}{r+1}(\mathbf{a}_0 + \dots + \mathbf{a}_r)$. Then \mathbf{a} lies in A and its barycentric coordinates $\lambda_0, \dots, \lambda_r$ are positive, each being $\frac{1}{r+1}$. By the continuity of the barycentric coordinates, for each $i = 0, \dots, r$, there exists $s_i > 0$, such that $\lambda_i > 0$ whenever \mathbf{x} lies in $B(\mathbf{a}; s_i) \cap \text{aff}A$. Let s be the minimum of s_0, \dots, s_r . So $s > 0$. Then if \mathbf{x} lies in $B(\mathbf{a}; s) \cap \text{aff}A$, all its barycentric coordinates $\lambda_0, \dots, \lambda_r$ are positive. So, since A is convex, \mathbf{x} lies in A . Thus, the relative interior of A contains the point \mathbf{a} .

Convex Sets with Empty Interior

Corollary

A convex set in \mathbb{R}^n has an empty interior if and only if it lies in some hyperplane of \mathbb{R}^n .

- Since a hyperplane in \mathbb{R}^n has an empty interior, so does each of its subsets.

A convex set in \mathbb{R}^n which does not lie in any hyperplane of \mathbb{R}^n must be n -dimensional. Therefore its interior coincides with its relative interior. This relative interior is non-empty by the theorem.

Line Segment Between Relative Interior and Set

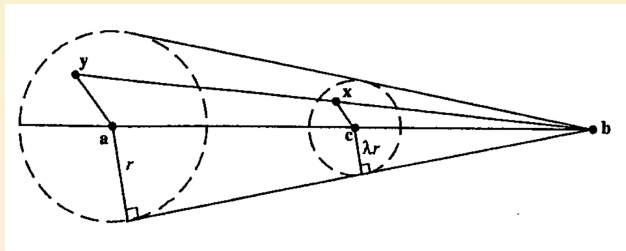
Lemma

Let A be a convex set in \mathbb{R}^n . Let $\mathbf{a} \in \text{ri}A$ and $\mathbf{b} \in A$. Then, for $0 < \lambda \leq 1$,

$$\lambda \mathbf{a} + (1 - \lambda) \mathbf{b} \in \text{ri}A.$$

- Since $\mathbf{a} \in \text{ri}A$, there is an $r > 0$ such that $B(\mathbf{a}; r) \cap \text{aff}A \subseteq A$.
Let $\mathbf{c} = \lambda \mathbf{a} + (1 - \lambda) \mathbf{b}$, where $0 < \lambda \leq 1$.
We show that $B(\mathbf{c}; \lambda r) \cap \text{aff}A \subseteq A$.

Line Segment Between Relative Interior and Set (Cont'd)



- Let $\mathbf{x} \in B(\mathbf{c}; \lambda r) \cap \text{aff} A$. Let

$$\mathbf{y} = \mathbf{a} + \frac{1}{\lambda}(\mathbf{x} - \mathbf{c}) = \mathbf{a} + \frac{1}{\lambda}(\mathbf{x} - \lambda \mathbf{a} - (1 - \lambda)\mathbf{b}) = \frac{1}{\lambda}\mathbf{x} + \left(1 - \frac{1}{\lambda}\right)\mathbf{b}.$$

Then $\mathbf{y} \in \text{aff} A$ and $\|\mathbf{y} - \mathbf{a}\| = \frac{1}{\lambda}\|\mathbf{x} - \mathbf{c}\| < r$. Thus, $\mathbf{y} \in B(\mathbf{a}; r) \cap \text{aff} A \subseteq A$. The equation $\mathbf{x} = \lambda \mathbf{y} + (1 - \lambda)\mathbf{b}$, together with the convexity of A , shows that $\mathbf{x} \in A$. Hence, $B(\mathbf{c}; \lambda r) \cap \text{aff} A \subseteq A$. So $\mathbf{c} \in \text{ri} A$.

A Closure and a Relative Interior Point of a Convex Set

Theorem

Let A be a convex set in \mathbb{R}^n . Let $\mathbf{a} \in \text{ri}A$ and $\mathbf{b} \in \text{cl}A$. Then $\lambda\mathbf{a} + (1-\lambda)\mathbf{b} \in \text{ri}A$ for $0 < \lambda \leq 1$.

- Since $\mathbf{a} \in \text{ri}A$, there is an $r > 0$ such that $B(\mathbf{a}; r) \cap \text{aff}A \subseteq A$. Let $\mathbf{c} = \lambda\mathbf{a} + (1-\lambda)\mathbf{b}$, where $0 < \lambda \leq 1$. Since $\mathbf{b} \in \text{cl}A$, there exists $\mathbf{d} \in A$ satisfying $(1-\lambda)\|\mathbf{d} - \mathbf{b}\| < \lambda r$. Let

$$\begin{aligned} \mathbf{e} &= \mathbf{a} + \frac{1-\lambda}{\lambda}(\mathbf{b} - \mathbf{d}) = \frac{1}{\lambda}(\lambda\mathbf{a} + (1-\lambda)\mathbf{b} - (1-\lambda)\mathbf{d}) \\ &= \frac{1}{\lambda}(\mathbf{c} + (\lambda - 1)\mathbf{d}) = \frac{1}{\lambda}\mathbf{c} + \left(1 - \frac{1}{\lambda}\right)\mathbf{d}. \end{aligned}$$

Then $\mathbf{e} \in \text{aff}A$ and $\|\mathbf{e} - \mathbf{a}\| = \frac{1-\lambda}{\lambda}\|\mathbf{b} - \mathbf{d}\| < r$. Thus \mathbf{e} lies in $B(\mathbf{a}; r) \cap \text{aff}A$. Hence, it lies in $\text{ri}A$. The equation $\mathbf{c} = \lambda\mathbf{e} + (1-\lambda)\mathbf{d}$, together with the lemma, shows that $\mathbf{c} \in \text{ri}A$.

Relative Interior, Interior and Closure of Convex Sets

Theorem

Let A be a convex set in \mathbb{R}^n . Then $\text{ri}A$, $\text{int}A$ and $\text{cl}A$ are convex.

- If $\mathbf{a}, \mathbf{b} \in \text{ri}A$ and $0 \leq \lambda \leq 1$, then $\lambda\mathbf{a} + (1-\lambda)\mathbf{b} \in \text{ri}A$, either trivially, if $\lambda = 0$, or by the preceding theorem, otherwise. Thus $\text{ri}A$ is convex.

That $\text{int}A$ is convex follows from the fact that either $\text{int}A$ is empty or coincides with $\text{ri}A$.

If $\mathbf{a}, \mathbf{b} \in \text{cl}A$, then there are sequences $\mathbf{a}_1, \dots, \mathbf{a}_k, \dots$ and $\mathbf{b}_1, \dots, \mathbf{b}_k, \dots$ of points of A such that $\mathbf{a}_k \rightarrow \mathbf{a}$, $\mathbf{b}_k \rightarrow \mathbf{b}$ as $k \rightarrow \infty$. Let $0 \leq \lambda \leq 1$. Then $\lambda\mathbf{a}_k + (1-\lambda)\mathbf{b}_k \in A$ for each k , since A is convex. Now

$$\lambda\mathbf{a}_k + (1-\lambda)\mathbf{b}_k \rightarrow \lambda\mathbf{a} + (1-\lambda)\mathbf{b} \text{ as } k \rightarrow \infty.$$

This shows that $\lambda\mathbf{a} + (1-\lambda)\mathbf{b} \in \text{cl}A$. Thus $\text{cl}A$ is convex.

The Closed Convex Hull of a Set

- The preceding theorem shows that, for any set A in \mathbb{R}^n , the set $\text{cl}(\text{conv}A)$ is a closed convex set containing A .
- If B is any closed convex set in \mathbb{R}^n containing A , then

$$B = \text{cl}(\text{conv}B) \supseteq \text{cl}(\text{conv}A).$$

- So $\text{cl}(\text{conv}A)$ is the smallest closed convex set containing A .
- It is called the **closed convex hull** of A .

Characterization of Relative Interior Points

- The next result asserts that a point \mathbf{a}_0 of a convex set A is a relative interior point of A if and only if every line segment lying in A , and having \mathbf{a}_0 as an endpoint, can be extended some distance beyond \mathbf{a}_0 without leaving A .

Theorem

Let \mathbf{a}_0 be a point of a convex set A in \mathbb{R}^n . Then $\mathbf{a}_0 \in \text{ri}A$ if and only if, for each $\mathbf{a} \in A$, there exists $\mu > 1$ such that $(1 - \mu)\mathbf{a} + \mu\mathbf{a}_0 \in A$.

- Clearly, if $\mathbf{a}_0 \in \text{ri}A$, then the condition of the theorem is satisfied. Conversely, suppose that \mathbf{a}_0 satisfies this condition. Let $\mathbf{a} \in \text{ri}A$. Then there is $\mu > 1$ such that the point $\mathbf{x} = (1 - \mu)\mathbf{a} + \mu\mathbf{a}_0$ lies in A . Hence $\mathbf{a}_0 = \lambda\mathbf{a} + (1 - \lambda)\mathbf{x}$, where $0 < \lambda = 1 - \frac{1}{\mu} < 1$. But $\mathbf{a} \in \text{ri}A$ and $\mathbf{x} \in A$, so $\mathbf{a}_0 \in \text{ri}A$ by a previous lemma.

Relative Interior of the Convex Hull of a Finite Set

Theorem

Let $A = \text{conv}\{\mathbf{a}_1, \dots, \mathbf{a}_m\}$, where $\mathbf{a}_1, \dots, \mathbf{a}_m \in \mathbb{R}^n$. Then

$$\text{ri}A = \{\lambda_1 \mathbf{a}_1 + \dots + \lambda_m \mathbf{a}_m : \lambda_1, \dots, \lambda_m > 0, \lambda_1 + \dots + \lambda_m = 1\}.$$

- Suppose first that $\mathbf{a}_0 = \lambda_1 \mathbf{a}_1 + \dots + \lambda_m \mathbf{a}_m$, where $\lambda_1, \dots, \lambda_m > 0$ and $\lambda_1 + \dots + \lambda_m = 1$, and that $\mathbf{a} \in A$. Then $\mathbf{a}_0 \in A$, and $\mathbf{a} = \mu_1 \mathbf{a}_1 + \dots + \mu_m \mathbf{a}_m$ for some $\mu_1, \dots, \mu_m \geq 0$ with $\mu_1 + \dots + \mu_m = 1$. Choose $\mu > 1$ such that

$$\mu \lambda_1 + (1 - \mu) \mu_1 \geq 0, \dots, \mu \lambda_m + (1 - \mu) \mu_m \geq 0.$$

Then $(1 - \mu)\mathbf{a} + \mu\mathbf{a}_0 \in A$. So $\mathbf{a}_0 \in \text{ri}A$ by the preceding theorem.

Relative Interior of the Convex Hull of a Finite Set (Cont'd)

- Suppose next that $\mathbf{a}_0 \in \text{ri}A$, and that $\mathbf{a}^* = \frac{1}{m}(\mathbf{a}_1 + \cdots + \mathbf{a}_m)$. Then $\mathbf{a}^* \in A$. By the preceding theorem, there exist $\mu > 1$ and $\mathbf{a} \in A$ such that $\mathbf{a} = (1 - \mu)\mathbf{a}^* + \mu\mathbf{a}_0$, say $\mathbf{a} = \mu_1\mathbf{a}_1 + \cdots + \mu_m\mathbf{a}_m$, where $\mu_1, \dots, \mu_m \geq 0$ with $\mu_1 + \cdots + \mu_m = 1$. The equation

$$\mathbf{a}_0 = \frac{\mu_1 + \frac{\mu-1}{m}}{\mu} \mathbf{a}_1 + \cdots + \frac{\mu_m + \frac{\mu-1}{m}}{\mu} \mathbf{a}_m$$

now expresses \mathbf{a}_0 in the form $\lambda_1\mathbf{a}_1 + \cdots + \lambda_m\mathbf{a}_m$, where $\lambda_1, \dots, \lambda_m > 0$ with $\lambda_1 + \cdots + \lambda_m = 1$.

Closure and Relative Interior

Theorem

Let A be a convex set in \mathbb{R}^n . Then $\text{ri}A = \text{ri}(\text{cl}A)$ and $\text{cl}A = \text{cl}(\text{ri}A)$.

- We assume, throughout, that A is non-empty with $\mathbf{a} \in \text{ri}A$.

The inclusion $\text{ri}A \subseteq \text{ri}(\text{cl}A)$ follows from the inclusion $A \subseteq \text{cl}A$ and the fact that the affine hulls of A and $\text{cl}A$ coincide. To establish the inclusion $\text{ri}(\text{cl}A) \subseteq \text{ri}A$, suppose that $\mathbf{b} \in \text{ri}(\text{cl}A)$. By a previous theorem, there exist $\mu > 1$ and $\mathbf{c} \in \text{cl}A$ such that $\mathbf{c} = (1 - \mu)\mathbf{a} + \mu\mathbf{b}$. Hence $\mathbf{b} = (1 - \lambda)\mathbf{a} + \lambda\mathbf{c}$, where $0 < \lambda = \frac{1}{\mu} < 1$. That $\mathbf{b} \in \text{ri}A$ follows from a previous theorem. Thus, $\text{ri}(\text{cl}A) \subseteq \text{ri}A$.

The inclusion $\text{cl}(\text{ri}A) \subseteq \text{cl}A$ is clear. To establish the inclusion $\text{cl}A \subseteq \text{cl}(\text{ri}A)$, suppose that $\mathbf{b} \in \text{cl}A$. A previous theorem shows that $\lambda\mathbf{a} + (1 - \lambda)\mathbf{b} \in \text{ri}A$ for $0 < \lambda \leq 1$. Hence, $\mathbf{b} \in \text{cl}(\text{ri}A)$. Thus, $\text{cl}A \subseteq \text{cl}(\text{ri}A)$.

Interior, Closure and Boundaries

Corollary

Let A be a convex set in \mathbb{R}^n . Then $\text{int}A = \text{int}(\text{cl}A)$ and, when $\text{int}A$ is nonempty, $\text{cl}A = \text{cl}(\text{int}A)$.

- If $\text{int}A$ is non-empty, then $\text{ri}A = \text{int}A$ and $\text{ri}(\text{cl}A) = \text{int}(\text{cl}A)$, and the corollary follows from the theorem.

If $\text{int}A$ is empty, then A , and hence $\text{cl}A$, lie in a hyperplane of \mathbb{R}^n . Hence, both $\text{int}A$ and $\text{int}(\text{cl}A)$ are empty.

Corollary

Let A be a convex set in \mathbb{R}^n . Then $\text{rebd}A = \text{rebd}(\text{cl}A)$ and $\text{bd}A = \text{bd}(\text{cl}A)$.

- By the theorem and its first corollary,

$$\text{rebd}(\text{cl}A) = \text{cl}(\text{cl}A) \setminus \text{ri}(\text{cl}A) = \text{cl}A \setminus \text{ri}A = \text{rebd}A;$$

$$\text{bd}(\text{cl}A) = \text{cl}(\text{cl}A) \setminus \text{int}(\text{cl}A) = \text{cl}A \setminus \text{int}A = \text{bd}A.$$

Subsection 4

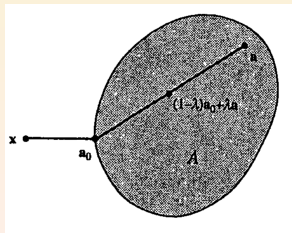
Separation and Support

Uniqueness of Closest Point

Theorem

In \mathbb{R}^n let A be a nonempty closed convex set and let \mathbf{x} be a point. Then there exists a unique point \mathbf{a}_0 of A such that $\|\mathbf{x} - \mathbf{a}_0\| = \inf \{\|\mathbf{x} - \mathbf{z}\| : \mathbf{z} \in A\}$. Moreover, $(\mathbf{x} - \mathbf{a}_0) \cdot (\mathbf{a} - \mathbf{a}_0) \leq 0$, for each \mathbf{a} in A .

- By a previous theorem, there exists $\mathbf{a}_0 \in A$ such that $\|\mathbf{x} - \mathbf{a}_0\| = \inf \{\|\mathbf{x} - \mathbf{z}\| : \mathbf{z} \in A\}$. Let $\mathbf{a} \in A$ and $0 < \lambda \leq 1$. The convexity of A shows $(1 - \lambda)\mathbf{a}_0 + \lambda\mathbf{a} \in A$.



The choice of \mathbf{a}_0 shows that

$$\|\mathbf{x} - ((1 - \lambda)\mathbf{a}_0 + \lambda\mathbf{a})\| = \|(\mathbf{x} - \mathbf{a}_0) + \lambda(\mathbf{a}_0 - \mathbf{a})\| \geq \|\mathbf{x} - \mathbf{a}_0\|.$$

We deduce, using a previous theorem that $(\mathbf{x} - \mathbf{a}_0) \cdot (\mathbf{a} - \mathbf{a}_0) \leq 0$.

Uniqueness of Closest Point (Cont'd)

- Suppose that $\mathbf{a}_1 \in A$ also satisfies the equation

$$\|\mathbf{x} - \mathbf{a}_1\| = \inf \{\|\mathbf{x} - \mathbf{z}\| : \mathbf{z} \in A\}.$$

Then, by what we have just proved, $(\mathbf{x} - \mathbf{a}_0) \cdot (\mathbf{a}_1 - \mathbf{a}_0) \leq 0$. Because of the symmetry between \mathbf{a}_0 and \mathbf{a}_1 , we have $(\mathbf{x} - \mathbf{a}_1) \cdot (\mathbf{a}_0 - \mathbf{a}_1) \leq 0$. Adding these last two inequalities together, we deduce that

$$\|\mathbf{a}_1 - \mathbf{a}_0\|^2 = (\mathbf{a}_1 - \mathbf{a}_0) \cdot (\mathbf{a}_1 - \mathbf{a}_0) \leq 0.$$

This proves that $\mathbf{a}_0 = \mathbf{a}_1$.

- The theorem shows how each non-empty closed convex set A in \mathbb{R}^n gives rise to a mapping $f : \mathbb{R}^n \rightarrow A$ defined by $f(\mathbf{x}) = \mathbf{a}_0$, where \mathbf{a}_0 is the nearest point of A to a point \mathbf{x} of \mathbb{R}^n .

This mapping is called the **projection operator** of A .

Lipschitz Property of Projection

Corollary

Let A be a non-empty closed convex set in \mathbb{R}^n . Then the projection operator $f : \mathbb{R}^n \rightarrow A$ of A satisfies the Lipschitz condition $\|f(\mathbf{x}) - f(\mathbf{y})\| \leq \|\mathbf{x} - \mathbf{y}\|$ for all $\mathbf{x}, \mathbf{y} \in \mathbb{R}^n$. So it is continuous.

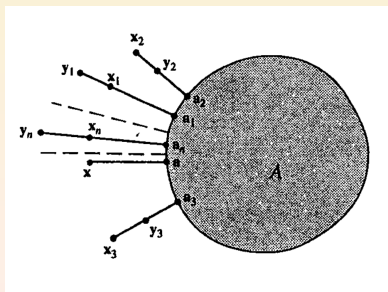
- Let $\mathbf{x}, \mathbf{y} \in \mathbb{R}^n$. Write $\mathbf{u} = \mathbf{x} - f(\mathbf{x})$, $\mathbf{v} = \mathbf{y} - f(\mathbf{y})$. Then, by the theorem, $\mathbf{u} \cdot (f(\mathbf{y}) - f(\mathbf{x})) \leq 0$ and $\mathbf{v} \cdot (f(\mathbf{x}) - f(\mathbf{y})) \leq 0$. So, we get $(\mathbf{u} - \mathbf{v}) \cdot (f(\mathbf{x}) - f(\mathbf{y})) \geq 0$. Thus,

$$\begin{aligned} \|\mathbf{x} - \mathbf{y}\|^2 &= \|(\mathbf{u} - \mathbf{v}) + (f(\mathbf{x}) - f(\mathbf{y}))\|^2 \\ &= \|\mathbf{u} - \mathbf{v}\|^2 + 2(\mathbf{u} - \mathbf{v}) \cdot (f(\mathbf{x}) - f(\mathbf{y})) + \|f(\mathbf{x}) - f(\mathbf{y})\|^2 \\ &\geq \|f(\mathbf{x}) - f(\mathbf{y})\|^2. \end{aligned}$$

So $\|f(\mathbf{x}) - f(\mathbf{y})\| \leq \|\mathbf{x} - \mathbf{y}\|$.

Geometry of Nearest Points

- Geometrically, the following corollary states that, if $f(x)$ is the nearest point of a non-empty closed convex set A in \mathbb{R}^n to a point x of \mathbb{R}^n not belonging to A ,



then it is also the nearest point of A to any point on the halfline starting at $f(x)$ and passing through x .

Nearest Points and Line Segments

Corollary

Let A be a non-empty closed convex set in \mathbb{R}^n with projection operator f . Then, for all $\mathbf{x} \in \mathbb{R}^n$ and $\lambda \geq 0$,

$$f(f(\mathbf{x}) + \lambda(\mathbf{x} - f(\mathbf{x}))) = f(\mathbf{x}).$$

- Write $\mathbf{y} = f(\mathbf{x}) + \lambda(\mathbf{x} - f(\mathbf{x}))$, where $\mathbf{x} \in \mathbb{R}^n$, $\lambda \geq 0$. By the theorem, $(\mathbf{x} - f(\mathbf{x})) \cdot (f(\mathbf{y}) - f(\mathbf{x})) \leq 0$ and $(\mathbf{y} - f(\mathbf{y})) \cdot (f(\mathbf{x}) - f(\mathbf{y})) \leq 0$.

From these inequalities, we deduce that

$$\begin{aligned} 0 &\leq (f(\mathbf{y}) - f(\mathbf{x})) \cdot (f(\mathbf{y}) - f(\mathbf{x})) \\ &= (f(\mathbf{y}) + \lambda(\mathbf{x} - f(\mathbf{x})) - \mathbf{y}) \cdot (f(\mathbf{y}) - f(\mathbf{x})) \\ &= (f(\mathbf{y}) - \mathbf{y}) \cdot (f(\mathbf{y}) - f(\mathbf{x})) + \lambda(\mathbf{x} - f(\mathbf{x})) \cdot (f(\mathbf{y}) - f(\mathbf{x})) \\ &\leq 0. \end{aligned}$$

Hence, $\|f(\mathbf{y}) - f(\mathbf{x})\|^2 = 0$. Therefore, $f(\mathbf{y}) = f(\mathbf{x})$.

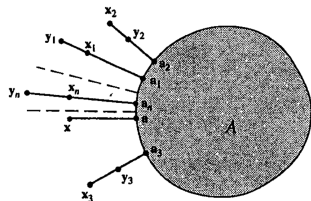
Relative Boundaries and Closest Points

Theorem

Let \mathbf{a} be a relative boundary point of a non-empty closed convex set A in \mathbb{R}^n with projection operator f . Then there exists $\mathbf{x} \in (\text{aff}A) \setminus A$ such that $f(\mathbf{x}) = \mathbf{a}$.

- Since $\mathbf{a} \in \text{rebd}A$, there exists, for each positive integer m , a point \mathbf{y}_m of $(\text{aff}A) \setminus A$ satisfying $\|\mathbf{y}_m - \mathbf{a}\| \leq \frac{1}{m}$. Write $\mathbf{a}_m = f(\mathbf{y}_m)$ and

$$\mathbf{x}_m = \mathbf{a}_m + \frac{\mathbf{y}_m - \mathbf{a}_m}{\|\mathbf{y}_m - \mathbf{a}_m\|}.$$



Relative Boundaries and Closest Points (Cont'd)

- Then $\|\mathbf{x}_m - \mathbf{a}_m\| = 1$ and, by the preceding corollary, $f(\mathbf{x}_m) = \mathbf{a}_m$. A previous corollary shows that

$$\|\mathbf{a}_m - \mathbf{a}\| = \|f(\mathbf{y}_m) - f(\mathbf{a})\| \leq \|\mathbf{y}_m - \mathbf{a}\| \leq \frac{1}{m}.$$

So $\mathbf{a}_m \rightarrow \mathbf{a}$ as $m \rightarrow \infty$. We have

$$\|\mathbf{x}_m\| \leq \|\mathbf{x}_m - \mathbf{a}_m\| + \|\mathbf{a}_m - \mathbf{a}\| + \|\mathbf{a}\| \leq 1 + \frac{1}{m} + \|\mathbf{a}\|.$$

So the sequence $\mathbf{x}_1, \mathbf{x}_2, \dots$ is bounded. Thus $\mathbf{x}_1, \mathbf{x}_2, \dots$ contains a convergent subsequence. Assume, without loss of generality, that $\mathbf{x}_1, \mathbf{x}_2, \dots$ itself converges to some point \mathbf{x} of \mathbb{R}^n . Since $\mathbf{x}_1, \mathbf{x}_2, \dots$ belong to $\text{aff}A$, we can deduce that $\mathbf{x} \in \text{aff}A$. The continuity of f shows that $f(\mathbf{x}_m) \rightarrow f(\mathbf{x})$ as $m \rightarrow \infty$, i.e., $\mathbf{a}_m \rightarrow f(\mathbf{x})$ as $m \rightarrow \infty$. But $\mathbf{a}_m \rightarrow \mathbf{a}$ as $m \rightarrow \infty$. Hence, $f(\mathbf{x}) = \mathbf{a}$. Clearly $\|\mathbf{x} - \mathbf{a}\| = 1$. So $\mathbf{x} \notin A$.

Bounded Sets and Boundaries

Corollary

Let A be a non-empty closed convex set in \mathbb{R}^n with projection operator f and let B be a bounded set in \mathbb{R}^n such that $A \subseteq B$. Then $f(\text{bd}B) = \text{bd}A$.

- We show that $\text{bd}A \subseteq f(\text{bd}B)$, the opposite inclusion being obvious.

Let $\mathbf{a} \in \text{bd}A$. Then there exists $\mathbf{x} \in \mathbb{R}^n \setminus A$ such that $f(\mathbf{x}) = \mathbf{a}$. This can be proved by substituting $\text{bd}A$ for $\text{rebd}A$, and \mathbb{R}^n for $\text{aff}A$, in the proof of the theorem. As B is bounded, there is some $\mu > 0$ such that, for $\lambda \geq \mu$, $\mathbf{a} + \lambda(\mathbf{x} - \mathbf{a}) \notin B$. But $\mathbf{a} \in B$. So $\mathbf{a} + \lambda_0(\mathbf{x} - \mathbf{a}) \in \text{bd}B$ for some $\lambda_0 \geq 0$. The preceding corollary shows that $f(\mathbf{a} + \lambda_0(\mathbf{x} - \mathbf{a})) = \mathbf{a}$. Hence, $\text{bd}A \subseteq f(\text{bd}B)$.

Separation

- Let A and B be sets, and let H be a hyperplane in \mathbb{R}^n .
- Then H is said to **separate** A and B if A lies in one of the closed halfspaces determined by H and B lies in the other.
- H is said to **separate** A and B **properly** if it separates them, but not both A and B lie in H .
- If A and B lie in opposite open halfspaces determined by H , then H is said to **separate** A and B **strictly**.
- It follows from the convexity of halfspaces that, if a hyperplane separates two sets, then it also separates their convex hulls.
For this reason, we consider only the separation of convex sets.

Examples

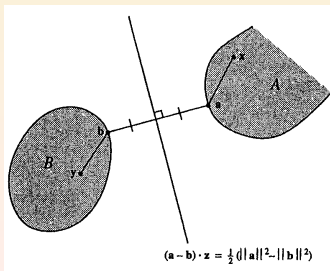
- It is not always possible to separate two convex sets by a hyperplane. For example, there is no line separating the set $\{0\}$ and the closed unit disc $\{(x, y) : x^2 + y^2 \leq 1\}$ in \mathbb{R}^2 .
- Any two sets that can be strictly separated can be properly separated, unless they are both empty.
- The convex sets $\{(x, y) : x \leq 0\}$ and $\{(x, y) : x > 0, y \geq \frac{1}{x}\}$ in \mathbb{R}^2 cannot be strictly separated, but they are properly separated by the y -axis.
- A hyperplane in \mathbb{R}^n separates any two of its subsets, but does not separate them properly.

Strict Separation of a Closed and a Compact Set

Theorem

Let A and B be disjoint non-empty convex sets in \mathbb{R}^n with A closed and B compact. Then A and B can be strictly separated by a hyperplane in \mathbb{R}^n .

- The geometry of the proof is as follows. Let \mathbf{a} and \mathbf{b} be nearest points of A and B . Then the hyperplane through the midpoint of the line segment joining \mathbf{a} and \mathbf{b} with normal vector $\mathbf{a} - \mathbf{b}$ strictly separates A and B .



Strict Separation of a Closed and a Compact Set (Cont'd)

- Let $\mathbf{a} \in A$, $\mathbf{b} \in B$ be such that \mathbf{a} is the nearest point of A to \mathbf{b} , and \mathbf{b} is the nearest point of B to \mathbf{a} . This is possible by a previous theorem. Since A and B are disjoint, $\mathbf{a} \neq \mathbf{b}$. Let $\mathbf{x} \in A$, $\mathbf{y} \in B$. Then, by a previous theorem, $(\mathbf{b} - \mathbf{a}) \cdot (\mathbf{x} - \mathbf{a}) \leq 0$ and $(\mathbf{a} - \mathbf{b}) \cdot (\mathbf{y} - \mathbf{b}) \leq 0$. Thus,

$$\begin{aligned}
 (\mathbf{a} - \mathbf{b}) \cdot \mathbf{x} &\geq (\mathbf{a} - \mathbf{b}) \cdot \mathbf{a} \\
 &= \frac{1}{2}(\|\mathbf{a}\|^2 - \|\mathbf{b}\|^2 + \|\mathbf{a} - \mathbf{b}\|^2) \\
 &> \frac{1}{2}(\|\mathbf{a}\|^2 - \|\mathbf{b}\|^2) \\
 &> \frac{1}{2}(\|\mathbf{a}\|^2 - \|\mathbf{b}\|^2 - \|\mathbf{a} - \mathbf{b}\|^2) \\
 &= (\mathbf{a} - \mathbf{b}) \cdot \mathbf{b} \\
 &\geq (\mathbf{a} - \mathbf{b}) \cdot \mathbf{y}.
 \end{aligned}$$

Write $\mathbf{c} = \mathbf{a} - \mathbf{b}$ and $c_0 = \frac{1}{2}(\|\mathbf{a}\|^2 - \|\mathbf{b}\|^2)$. Then we have shown that the hyperplane $\mathbf{c} \cdot \mathbf{z} = c_0$ strictly separates A and B .

Consequences

Corollary

In \mathbb{R}^n let A be a closed convex set and let \mathbf{b} be a point not lying in A . Then A and $\{\mathbf{b}\}$ can be strictly separated by a hyperplane in \mathbb{R}^n .

Corollary

Each closed convex set A in \mathbb{R}^n is the intersection of all the closed halfspaces in \mathbb{R}^n containing A .

- Denote by B the intersection of all the closed halfspaces in \mathbb{R}^n containing A . Then B is a closed convex set containing A . If $\mathbf{b} \notin A$, then the corollary above shows that there exists some closed halfspace in \mathbb{R}^n which contains A but not \mathbf{b} . Hence $\mathbf{b} \notin B$. Thus $B \subseteq A$, and $A = B$.

Convex Sets Not Containing the Origin

Lemma

In \mathbb{R}^n let A be a non-empty convex set not containing the origin. Then there exists a hyperplane in \mathbb{R}^n which separates the origin and A , and does not contain A .

- Suppose first that $\mathbf{0} \notin \text{cl}A$. Then the lemma follows from a previous corollary applied to the closed convex set $\text{cl}A$. Suppose next that $\mathbf{0} \in \text{cl}A$. Then $\mathbf{0} \in \text{rebd}A$. By a previous corollary, $\text{rebd}A = \text{rebd}(\text{cl}A)$. So $\mathbf{0} \in \text{rebd}(\text{cl}A)$. A previous theorem asserts the existence of a point \mathbf{x} of $(\text{aff}(\text{cl}A)) \setminus \text{cl}A$ whose nearest point in $\text{cl}A$ is $\mathbf{0}$. By a previous theorem, $\text{cl}A \subseteq \{\mathbf{z} \in \mathbb{R}^n : \mathbf{x} \cdot \mathbf{z} \leq 0\}$. This shows that the hyperplane $H = \{\mathbf{z} \in \mathbb{R}^n : \mathbf{x} \cdot \mathbf{z} = 0\}$ separates $\mathbf{0}$ and A .

We cannot have $A \subseteq H$: This would imply that $\mathbf{x} \in \text{aff}(\text{cl}A) \subseteq H$. But, this is impossible since $\mathbf{x} \cdot \mathbf{x} > 0$. Thus, H separates $\{\mathbf{0}\}$ and A , and does not contain A .

Disjoint Nonempty Convex Sets

Theorem

Each pair of disjoint non-empty convex sets A and B in \mathbb{R}^n can be properly separated by a hyperplane in \mathbb{R}^n .

- The non-empty convex set $A - B$ does not contain the origin. By the lemma, there exists a hyperplane in \mathbb{R}^n which separates $\{\mathbf{0}\}$ and $A - B$, and that does not contain $A - B$. Thus, there exist $\mathbf{c} \in \mathbb{R}^n$ with $\mathbf{c} \neq \mathbf{0}$ and $c_0 \in \mathbb{R}$ such that

$$0 = \mathbf{c} \cdot \mathbf{0} \leq c_0 \quad \text{and} \quad \mathbf{c} \cdot (\mathbf{a} - \mathbf{b}) \geq c_0, \quad \text{for } \mathbf{a} \in A, \mathbf{b} \in B.$$

Also, for some $\mathbf{a}_0 \in A$, $\mathbf{b}_0 \in B$, we have $\mathbf{c} \cdot (\mathbf{a}_0 - \mathbf{b}_0) > c_0 \geq 0$.

For every $\mathbf{a} \in A$, $\mathbf{b} \in B$,

$$\mathbf{c} \cdot \mathbf{a} \geq \mathbf{c} \cdot \mathbf{b} + c_0 \geq \mathbf{c} \cdot \mathbf{b}.$$

Disjoint Nonempty Convex Sets (Cont'd)

- Thus there is a scalar d satisfying the inequalities

$$\inf \{ \mathbf{c} \cdot \mathbf{a} : \mathbf{a} \in A \} \geq d \geq \sup \{ \mathbf{c} \cdot \mathbf{b} : \mathbf{b} \in B \}.$$

For any $\mathbf{a}' \in A$, $\mathbf{b}' \in B$, we have $\mathbf{c} \cdot \mathbf{a}' \geq d \geq \mathbf{c} \cdot \mathbf{b}'$. So the hyperplane H with equation $\mathbf{c} \cdot \mathbf{z} = d$ separates A and B .

H cannot contain both A and B , for this would imply that $\mathbf{c} \cdot (\mathbf{a}_0 - \mathbf{b}_0) = 0$, which contradicts the inequality above. Thus H separates A and B properly.

Convex Sets With Disjoint Relative Interiors

Corollary

Each pair of non-empty convex sets A and B in \mathbb{R}^n whose relative interiors are disjoint can be properly separated by a hyperplane in \mathbb{R}^n .

- The non-empty convex sets $\text{ri}A$ and $\text{ri}B$ are disjoint. So, by the theorem, there exists a hyperplane H in \mathbb{R}^n which properly separates them. Since closed halfspaces are closed, H also properly separates $\text{cl}(\text{ri}A) = \text{cl}A$ and $\text{cl}(\text{ri}B) = \text{cl}B$. Hence, it also properly separates A and B .

Support Hyperplanes

- In \mathbb{R}^n a hyperplane H is called a **support hyperplane** to a set A if H meets $\text{cl}A$ and A lies in one of the closed halfspaces determined by H .
- Such a hyperplane H is said to **support** A at those points where H meets $\text{cl}A$.
- A hyperplane H cannot support a set A at an interior point of A , because every ball with center in H meets both the open halfspaces determined by H .
- A hyperplane in \mathbb{R}^n is a **trivial support hyperplane** to each of its non-empty subsets.
- A support hyperplane to a set in \mathbb{R}^n is said to be a **non-trivial support hyperplane** to the set if it does not contain the set itself.

Boundary Points and Support Hyperplanes

Theorem

Through each boundary point of a convex set A in \mathbb{R}^n there passes a support hyperplane to A , and through each relative boundary point of A there passes a non-trivial support hyperplane to A .

- Suppose first that \mathbf{a} is a boundary point of A , but not a relative boundary point of A . Then A cannot be n -dimensional. So it lies in some hyperplane H of \mathbb{R}^n . Clearly H is a support hyperplane to A passing through \mathbf{a} .

Boundary Points and Support Hyperplanes (Cont'd)

- Suppose next that \mathbf{a} is a relative boundary point of A .

The preceding theorem shows the existence of a hyperplane H which properly separates $\{\mathbf{a}\}$ and $\text{ri}A$.

H cannot contain $\text{ri}A$, for this would imply that H also contains $\text{cl}(\text{ri}A) = \text{cl}A$, and hence \mathbf{a} .

By the definition of separation, $\{\mathbf{a}\}$ and $\text{ri}A$, and thus $\{\mathbf{a}\}$ and $\text{cl}(\text{ri}A) = \text{cl}A$, belong to opposite closed halfspaces determined by H .

Since $\mathbf{a} \in \text{cl}A$, we must have $\mathbf{a} \in H$. Thus H is a non-trivial support hyperplane to A passing through \mathbf{a} .

Example

- Let H be a support hyperplane to a closed ball $B[\mathbf{a}; r]$ in \mathbb{R}^n at some point \mathbf{c} .
- Since H does not meet the interior of the ball, every point of H must be a distance of at least r from \mathbf{a} .
- Hence \mathbf{c} must be a nearest point of H to \mathbf{a} .
- By the uniqueness of nearest points of convex sets, H can meet the ball only in the point \mathbf{c} .
- A previous theorem shows that $(\mathbf{h} - \mathbf{c}) \cdot (\mathbf{a} - \mathbf{c}) \leq 0$ for all \mathbf{h} in H .
- We cannot have $(\mathbf{h} - \mathbf{c}) \cdot (\mathbf{a} - \mathbf{c}) < 0$ for some \mathbf{h} in H .
This would imply that the point $\mathbf{h}' = 2\mathbf{c} - \mathbf{h}$ of H satisfies

$$(\mathbf{h}' - \mathbf{c}) \cdot (\mathbf{a} - \mathbf{c}) = (2\mathbf{c} - \mathbf{h} - \mathbf{c}) \cdot (\mathbf{a} - \mathbf{c}) = (\mathbf{c} - \mathbf{h}) \cdot (\mathbf{a} - \mathbf{c}) > 0.$$

But this is impossible.

- Thus H must be the hyperplane with equation $(\mathbf{x} - \mathbf{c}) \cdot (\mathbf{a} - \mathbf{c}) = 0$.
- So H is the unique support hyperplane to $B[\mathbf{a}; r]$ at \mathbf{c} .

Distance Between a Point and a Hyperplane

- We conclude this section by establishing a formula for the distance between a point \mathbf{a} and a hyperplane H with equation $\mathbf{c} \cdot \mathbf{x} = c_0$ in \mathbb{R}^n .
- Denote by \mathbf{a}_0 the point defined by the equation

$$\mathbf{a}_0 = \mathbf{a} + \frac{c_0 - \mathbf{c} \cdot \mathbf{a}}{\|\mathbf{c}\|^2} \mathbf{c}.$$

- Then \mathbf{a}_0 lies in H , and for any \mathbf{x} in H , we have

$$\|\mathbf{a} - \mathbf{x}\|^2 = \|(\mathbf{a} - \mathbf{a}_0) + (\mathbf{a}_0 - \mathbf{x})\|^2 = \|\mathbf{a} - \mathbf{a}_0\|^2 + \|\mathbf{a}_0 - \mathbf{x}\|^2.$$

This shows that \mathbf{a}_0 is the unique nearest point of H to \mathbf{a} , and that the (shortest) distance between \mathbf{a} and H is $\|\mathbf{a} - \mathbf{a}_0\| = \frac{|\mathbf{c} \cdot \mathbf{a} - c_0|}{\|\mathbf{c}\|}$.

- When \mathbf{c} is a unit vector and \mathbf{a} is the origin, this distance becomes $|c_0|$.
- The (shortest) distance between parallel hyperplanes $\mathbf{c} \cdot \mathbf{x} = c_0$ and $\mathbf{c} \cdot \mathbf{x} = d_0$ is $\frac{|d_0 - c_0|}{\|\mathbf{c}\|}$ which becomes $|d_0 - c_0|$ when \mathbf{c} is a unit vector.

Subsection 5

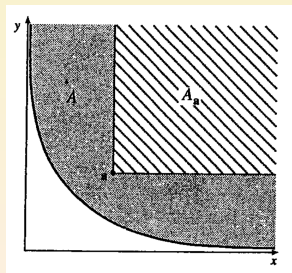
Unbounded Convex Sets

Example: Recession Cone

- Let A be the closed unbounded convex set in \mathbb{R}^2 that is defined by the equation

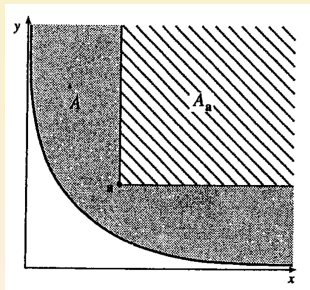
$$A = \{(x, y) : y \geq \frac{1}{x}, x > 0\}$$

and let $\mathbf{a} \in A$.



- Then there are halflines starting at \mathbf{a} which are contained in A .
- If we denote by $A_{\mathbf{a}}$ the union of all these halflines, then $A_{\mathbf{a}}$ is a closed convex set that is a union of halflines starting at \mathbf{a} .
- Such a set $A_{\mathbf{a}}$ is called a **closed convex cone with apex \mathbf{a}** .
- In fact, $A_{\mathbf{a}} = \mathbf{a} + P$, where P is the non-negative quadrant $\{(x, y) : x \geq 0, y \geq 0\}$ of \mathbb{R}^2 .

Recession Cone



- $A_{\mathbf{a}} = \mathbf{a} + P$, where $P = \{(x, y) : x \geq 0, y \geq 0\}$.
- The important observation here is that P is determined by the set A alone, being independent of the initial choice of the point \mathbf{a} in A .
- We refer to P as the **recession cone** of A .
- Roughly speaking, the recession cone of a convex set indicates in which directions the set recedes to infinity.

Halflines

- A **halfline** L^+ in \mathbb{R}^n is a set of the form $\{\mathbf{x}_0 + \lambda \mathbf{y} : \lambda \geq 0\}$, where $\mathbf{x}_0, \mathbf{y} \in \mathbb{R}^n$ and $\mathbf{y} \neq \mathbf{0}$.
- The reason for this is that the line joining the points \mathbf{x}_0 and $\mathbf{x}_0 + \mathbf{y}$ is the set

$$\{(1 - \lambda)\mathbf{x}_0 + \lambda(\mathbf{x}_0 + \mathbf{y}) : \lambda \in \mathbb{R}\} = \{\mathbf{x}_0 + \lambda \mathbf{y} : \lambda \in \mathbb{R}\}.$$

- A halfline L_0^+ of the form $\{\lambda \mathbf{y} : \lambda \geq 0\}$, where $\mathbf{y} \in \mathbb{R}^n$ and $\mathbf{y} \neq \mathbf{0}$, is called a **ray**.
- The equation $L^+ = \mathbf{x}_0 + L_0^+$ expresses L^+ as a translate of the ray L_0^+ .
- Since \mathbf{x}_0 is the only point of L^+ whose removal from L^+ leaves a convex set, and $L_0^+ = L^+ - \mathbf{x}_0$, it follows that \mathbf{x}_0 and L_0^+ are uniquely determined by L^+ .
- L^+ is the **halfline with direction** L_0^+ and **initial point** \mathbf{x}_0 .
- The word **direction** will be used as a synonym for ray.

Closed Unbounded Convex Sets and Halflines

Theorem

Let A be a closed unbounded convex set in \mathbb{R}^n . Then A contains a halfline. Moreover, if A contains some halfline with direction L_0^+ , then it contains every halfline with direction L_0^+ whose initial point is in A .

- Since A is unbounded, it contains a sequence $\mathbf{a}_1, \dots, \mathbf{a}_k, \dots$ of non-zero vectors such that $\|\mathbf{a}_k\| \rightarrow \infty$ as $k \rightarrow \infty$. Let $\lambda_k = \frac{1}{\|\mathbf{a}_k\|}$. Then the sequence $\lambda_1 \mathbf{a}_1, \dots, \lambda_k \mathbf{a}_k, \dots$ lies in the compact set $\{\mathbf{x} : \|\mathbf{x}\| = 1\}$. So it contains some subsequence converging to a point \mathbf{a} with $\|\mathbf{a}\| = 1$. We may suppose that the sequence itself converges to \mathbf{a} . Let L^+ be the direction $\{\lambda \mathbf{a} : \lambda \geq 0\}$. Then we show that $\mathbf{a}_0 + L^+ \subseteq A$ for every \mathbf{a}_0 in A .

Closed Unbounded Convex Sets and Halflines (Cont'd)

- Let $\mathbf{a}_0 \in A$ and let $\lambda \geq 0$. Since $\lambda_k \rightarrow 0$ as $k \rightarrow \infty$, we must have, for all but a finite number of k 's, $0 \leq \lambda \lambda_k \leq 1$ and $(1 - \lambda \lambda_k) \mathbf{a}_0 + \lambda \lambda_k \mathbf{a}_k \in A$. Clearly,

$$(1 - \lambda \lambda_k) \mathbf{a}_0 + \lambda \lambda_k \mathbf{a}_k \rightarrow \mathbf{a}_0 + \lambda \mathbf{a}, \text{ as } k \rightarrow \infty.$$

So $\mathbf{a}_0 + \lambda \mathbf{a} \in A$, since A is closed. Thus the halfline $\mathbf{a}_0 + L^+$ is contained in A .

Suppose next that A contains the halfline $\mathbf{b}_0 + L_0^+$, where L_0^+ is the direction $\{\lambda \mathbf{b} : \lambda \geq 0\}$ for some $\mathbf{b} \neq \mathbf{0}$. We show that every halfline $\mathbf{c}_0 + L_0^+$, where $\mathbf{c}_0 \in A$, is contained in A . Let $\mu \geq 0$. Then, for all $\lambda > \mu$,

$$\left(1 - \frac{\mu}{\lambda}\right) \mathbf{c}_0 + \frac{\mu}{\lambda} (\mathbf{b}_0 + \lambda \mathbf{b}) \in A.$$

Letting $\lambda \rightarrow \infty$ in this last relation and using the fact that A is closed, we deduce that $\mathbf{c}_0 + \mu \mathbf{b} \in A$. Thus $\mathbf{c}_0 + L_0^+ \subseteq A$.

Relative Interior, Closure and Halflines

Corollary

Let A be an unbounded convex set in \mathbb{R}^n . Then $\text{ri}A$ contains a halfline. Moreover, if $\text{cl}A$ contains some halfline with direction L_0^+ , then $\text{ri}A$ contains every halfline with direction L_0^+ whose initial point is in $\text{ri}A$.

- We apply the theorem to the closed unbounded convex set $\text{cl}A$.

Suppose that $\text{cl}A$ contains the halfline $\mathbf{b}_0 + L_0^+$, where L_0^+ is the direction $\{\lambda \mathbf{b} : \lambda \geq 0\}$. Let $\mathbf{a}_0 \in \text{ri}A$ and let $\mu \geq 0$. Then $\mathbf{a}_0 + 2\mu \mathbf{b} \in \text{cl}A$. So $\mathbf{a}_0 + \mu \mathbf{b} = \frac{1}{2}\mathbf{a}_0 + \frac{1}{2}(\mathbf{a}_0 + 2\mu \mathbf{b}) \in \text{ri}A$, by a previous theorem. Thus $\mathbf{a}_0 + L_0^+ \subseteq \text{ri}A$.

Cones

- A nonempty set A in \mathbb{R}^n is called a **cone** if $\lambda \mathbf{a} \in A$ whenever $\mathbf{a} \in A$ and $\lambda \geq 0$.
- Examples of cones are:
 - subspaces;
 - rays;
 - the **nonnegative orthant**

$$\{(x_1, \dots, x_n) : x_1 \geq 0, \dots, x_n \geq 0\}$$

of \mathbb{R}^n .

- All cones contain the origin and are, with the exception of the trivial cone $\{\mathbf{0}\}$, unbounded.
- Cones need not be convex.

The set $\{(x, y) : xy \geq 0\}$ is a non-convex cone in \mathbb{R}^2 .

Characterization of Convex Cones

Theorem

Let A be a non-empty set in \mathbb{R}^n . Then A is a convex cone if and only if $\lambda \mathbf{a} + \mu \mathbf{b} \in A$ whenever $\mathbf{a}, \mathbf{b} \in A$ and $\lambda, \mu \geq 0$.

- Let A be a convex cone. Suppose that $\mathbf{a}, \mathbf{b} \in A$ and $\lambda, \mu \geq 0$. If $\lambda + \mu = 0$, then $\lambda = \mu = 0$ and trivially $\lambda \mathbf{a} + \mu \mathbf{b} \in A$. If $\lambda + \mu > 0$, then

$$\frac{\lambda}{\lambda + \mu} \mathbf{a} + \frac{\mu}{\lambda + \mu} \mathbf{b} \in A,$$

since A is convex. Hence

$$\lambda \mathbf{a} + \mu \mathbf{b} = (\lambda + \mu) \left(\frac{\lambda}{\lambda + \mu} \mathbf{a} + \frac{\mu}{\lambda + \mu} \mathbf{b} \right) \in A,$$

since A is a cone. Thus, in all cases, $\lambda \mathbf{a} + \mu \mathbf{b} \in A$.

Characterization of Convex Cones (Cont'd)

- Suppose next that $\lambda \mathbf{a} + \mu \mathbf{b} \in A$ whenever $\mathbf{a}, \mathbf{b} \in A$ and $\lambda, \mu \geq 0$.

Clearly A is convex.

To show that A is a cone, let $\mathbf{a} \in A$ and $\lambda \geq 0$. Then, by our hypothesis, $\lambda \mathbf{a} = \lambda \mathbf{a} + \mathbf{0} \in A$.

Corollary

Let A be a non-empty set in \mathbb{R}^n . Then A is a convex cone if and only if $\mathbf{a} + \mathbf{b} \in A$ and $\lambda \mathbf{a} \in A$ whenever $\mathbf{a}, \mathbf{b} \in A$ and $\lambda \geq 0$.

Convex Cone Generated by a Set

- It is a routine matter to show that the intersection of any family of convex cones in \mathbb{R}^n is a convex cone.
- Hence $\text{cone}A$, defined as the intersection of all convex cones containing a set A in \mathbb{R}^n , is a convex cone.
- It is called the **convex cone generated by A** .
- Clearly $\text{cone}A$ is the smallest convex cone containing A .
- We note that $\text{cone}\emptyset = \{\mathbf{0}\}$.
- We now characterize $\text{cone}A$, in the case when A is non-empty, as the set of all **nonnegative linear combinations** of points of A , i.e., points of the form

$$\lambda_1 \mathbf{a}_1 + \cdots + \lambda_m \mathbf{a}_m,$$

where $\mathbf{a}_1, \dots, \mathbf{a}_m \in A$ and $\lambda_1, \dots, \lambda_m \geq 0$.

Characterization of Generated Convex Cones

Theorem

Let A be a nonempty set in \mathbb{R}^n . Then $\text{cone}A$ is the set of all non-negative linear combinations of points of A .

- Denote by B the set of all nonnegative linear combinations of points of A . Let $\mathbf{x} \in B$. Then

$$\mathbf{x} = \lambda_1 \mathbf{a}_1 + \cdots + \lambda_m \mathbf{a}_m, \text{ for some } \mathbf{a}_1, \dots, \mathbf{a}_m \in A \text{ and } \lambda_1, \dots, \lambda_m \geq 0.$$

Then $\mathbf{x} \in \text{cone}A$ by repeated use of the corollary to $\text{cone}A$. Hence $B \subseteq \text{cone}A$. The corollary shows that B is a convex cone. Clearly $A \subseteq B$. So $\text{cone}A \subseteq B$. Thus, $B = \text{cone}A$.

Recession Cone

- A nonempty convex set A in \mathbb{R}^n is said to **recede in a direction** L_0^+ , or to have a **direction of recession** L_0^+ , if every halfline with initial point in A and direction L_0^+ lies in A , i.e., if $A + L_0^+ \subseteq A$.
- The union of all directions of recession of A , together with the zero vector, is called the **recession cone** of A .
- A previous theorem shows that a nonempty closed convex set in \mathbb{R}^n is bounded if and only if its recession cone consists of the zero vector alone.
- The recession cone of a non-empty flat is the unique subspace which is parallel to it.
- The set $\{(x, y) : x > 0, y > 0\} \cup \{(0, 0)\}$ is its own recession cone. It is an example of a set whose recession cone is not closed.

Characterization of Recession Cone

Theorem

Let A be a non-empty convex set in \mathbb{R}^n . Then the recession cone of A consists of all those points \mathbf{x} such that $A + \mathbf{x} \subseteq A$. Moreover, the recession cone of A is a convex cone, which is closed when A is closed.

- If \mathbf{x} belongs to the recession cone of A , then trivially $A + \mathbf{x} \subseteq A$.
Conversely, if $A + \mathbf{x} \subseteq A$, then

$$A + 2\mathbf{x} = (A + \mathbf{x}) + \mathbf{x} \subseteq A + \mathbf{x} \subseteq A.$$

By repeated application of this argument, $A + m\mathbf{x} \subseteq A$ for each positive integer m . But A is convex. So $A + \lambda\mathbf{x} \subseteq A$, for all $\lambda \geq 0$. Hence \mathbf{x} lies in the recession cone of A .

Characterization of Recession Cone (Cont'd)

- Denote by C the recession cone of A . Let $\mathbf{x}, \mathbf{y} \in C$ and $\lambda, \mu \geq 0$. Then

$$A + \lambda\mathbf{x} + \mu\mathbf{y} = (A + \lambda\mathbf{x}) + \mu\mathbf{y} \subseteq A + \mu\mathbf{y} \subseteq A.$$

So $\lambda\mathbf{x} + \mu\mathbf{y} \in C$. Hence C is a convex cone by a previous theorem.

Suppose now that A is closed. Let $\mathbf{x}_1, \dots, \mathbf{x}_k, \dots$ be a sequence of points of the recession cone C that converges to some point \mathbf{x} of \mathbb{R}^n . Then $\mathbf{a} + \mathbf{x}_k \in A$ for each k and for each point \mathbf{a} of A . But A is closed, so $\mathbf{a} + \mathbf{x} \in A$ for each point \mathbf{a} of A . I.e., $A + \mathbf{x} \subseteq A$. Thus, $\mathbf{x} \in C$. This shows that C is closed.

Direction of a Line

- Let L be a line in \mathbb{R}^n .
- Then by the **direction** of L is meant the *unique* line L_0 in \mathbb{R}^n , which is parallel to L and passes through the origin.
- A line is uniquely determined by specifying one of its points and giving its direction.

Indeed, if \mathbf{x} lies on a line L in \mathbb{R}^n , then:

- The direction L_0 of L is simply the line $L - \mathbf{x}$;
- $L = \mathbf{x} + L_0$.

Lineality Space

- Let L_0 be a line in \mathbb{R}^n that passes through the origin.
- Then a nonempty convex set A in \mathbb{R}^n is said to be **linear in the direction** L_0 , or to have a **direction of linearity** L_0 , if every line meeting A which has direction L_0 lies in A , i.e., if $A + L_0 \subseteq A$.
- The union of all the directions of linearity of A , together with the zero vector, is called the **lineality space** of A .
- A previous theorem shows that, if a closed convex set A contains a line with direction L_0 , then it contains every line with direction L_0 which meets A , i.e., L_0 is a direction of linearity of A .

Lineality Space: Examples

- A non-empty closed convex set contains a line if and only if its lineality space does not consist of the zero vector alone.
- The lineality space of a non-empty flat is the unique subspace which is parallel to it.
- The lineality space of the unbounded circular cylinder

$$\{(x, y, z) : x^2 + y^2 \leq 1\}$$

is the subspace $\{(0, 0, z) : z \in \mathbb{R}\}$, i.e., the z -axis.

Characterization of Lineality Space

Theorem

Let A be a non-empty convex set in \mathbb{R}^n . Then the lineality space of A consists of all those points \mathbf{x} of \mathbb{R}^n such that $A + \mathbf{x} = A$, and is a subspace of \mathbb{R}^n .

- If \mathbf{x} belongs to the lineality space of A , then trivially $A + \mathbf{x} \subseteq A$ and $A - \mathbf{x} \subseteq A$. Hence, $A + \mathbf{x} = A$.

Conversely, if $A + \mathbf{x} = A$, then, as in the proof of the preceding theorem, $A + m\mathbf{x} = A$, for each positive integer m . If m is a negative integer, then $A + m\mathbf{x} = (A + (-m)\mathbf{x}) + m\mathbf{x} = A$. Hence $A + m\mathbf{x} = A$, for all integers m . But A is convex, so $A + \lambda\mathbf{x} = A$ for all real λ . Thus \mathbf{x} belongs to the lineality space of A .

Let S be the lineality space of A . Let $\mathbf{x}, \mathbf{y} \in S$ and $\lambda, \mu \in \mathbb{R}$. Then trivially $A + \lambda\mathbf{x} = A$ and $A + \mu\mathbf{y} = A$. Thus

$$A + \lambda\mathbf{x} + \mu\mathbf{y} = (A + \lambda\mathbf{x}) + \mu\mathbf{y} = A + \mu\mathbf{y} = A.$$

So $\lambda\mathbf{x} + \mu\mathbf{y} \in S$. This shows that S is a subspace of \mathbb{R}^n .

Decomposition of a Closed Convex Set

Theorem

Let A be a non-empty closed convex set in \mathbb{R}^n with lineality space S . Then

$$A = S + (A \cap S^\perp),$$

and the convex set $A \cap S^\perp$ contains no lines.

- Let $\mathbf{a} \in A$. Then \mathbf{a} can be expressed uniquely in the form $\mathbf{a} = \mathbf{b} + \mathbf{c}$, where $\mathbf{b} \in S$ and $\mathbf{c} \in S^\perp$. Since $\mathbf{b} \in S$, $-\mathbf{b} \in S$. Hence, by the preceding theorem, $\mathbf{c} = \mathbf{a} - \mathbf{b} \in A$. Thus $\mathbf{c} \in A \cap S^\perp$. So $A \subseteq S + (A \cap S^\perp)$. The opposite inclusion follows immediately from the preceding theorem. Thus $A = S + (A \cap S^\perp)$.

Suppose that $A \cap S^\perp$ does contain a line. Then there exist \mathbf{x}, \mathbf{y} in \mathbb{R}^n with $\mathbf{y} \neq \mathbf{0}$ such that $\mathbf{x} + \lambda \mathbf{y} \in A \cap S^\perp$ for all real λ . A previous theorem shows that $\mathbf{y} \in S$. Hence, for all real λ , $(\mathbf{x} + \lambda \mathbf{y}) \cdot \mathbf{y} = \mathbf{x} \cdot \mathbf{y} + \lambda \|\mathbf{y}\|^2 = 0$, which is clearly impossible. Thus $A \cap S^\perp$ contains no lines.

Subsection 6

Facial Structure

Faces of a Convex Set

- Each face F of a three-dimensional convex polyhedron P is a convex subset of P with the property that whenever the relative interior of a line segment L lying in P meets F , then the endpoints of L lie in F .
- This observation motivates the definition of a face of a general convex set.
- A **face** of a convex set A in \mathbb{R}^n is a convex subset B of A such that whenever $\lambda \mathbf{x} + \mu \mathbf{y} \in B$, where $\mathbf{x}, \mathbf{y} \in A$ and $\lambda, \mu > 0$ with $\lambda + \mu = 1$, then $\mathbf{x}, \mathbf{y} \in B$.
- Every convex set A in \mathbb{R}^n has the faces \emptyset and A , called **improper faces** of A .
- Faces of A other than \emptyset and A are called **proper faces** of A .

Example and k -Faces

- The above definition of a face is more comprehensive than the one usually understood in elementary geometry.

Example: A cube has:

- six two-dimensional faces;
 - one three-dimensional face (itself);
 - twelve one-dimensional faces (its edges);
 - eight zero-dimensional faces (its vertices);
 - one face of dimension -1 (the empty set).
- In general, we refer to a k -dimensional face of a convex set as a **k -face**.

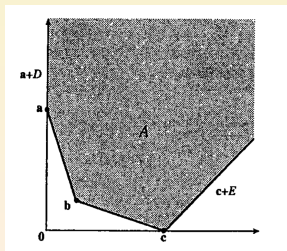
Special Names for Particular Faces

- Certain faces of a convex set are of particular importance and are given special names.
- The 0-faces of a convex set are called its **extreme points**.
- The faces that are halflines are called its **extreme half lines**.
- The directions of the extreme halflines of a convex set are called its **extreme directions**.
- Clearly, a point \mathbf{a} of a convex set A in \mathbb{R}^n is an extreme point of A if and only if whenever $\mathbf{a} = \lambda \mathbf{x} + \mu \mathbf{y}$, where $\mathbf{x}, \mathbf{y} \in A$ and $\lambda, \mu > 0$ with $\lambda + \mu = 1$, then $\mathbf{x} = \mathbf{y} = \mathbf{a}$.

Example

- Denote by A the set of those points (x, y) in \mathbb{R}^2 which satisfy the four inequalities

$$\begin{aligned} x &\geq 0, \\ -x + y + 1 &\geq 0, \\ x + 3y - 1 &\geq 0, \\ 3x + y - 1 &\geq 0. \end{aligned}$$



- The extreme points of A are the points $\mathbf{a} = (0, 1)$, $\mathbf{b} = (\frac{1}{4}, \frac{1}{4})$ and $\mathbf{c} = (1, 0)$.
- The extreme directions of A are the directions $D = \{(0, \lambda) : \lambda \geq 0\}$ and $E = \{(\lambda, \lambda) : \lambda \geq 0\}$.
- The extreme halflines of A are the halflines $\mathbf{a} + D$ and $\mathbf{c} + E$.

Characterization of the Faces

- Consider the case of a 2-face B of a cube A in \mathbb{R}^3 .
 - The set $A \setminus B$, i.e., the cube A with its face B removed, is convex.
 - Also B is the intersection of its affine hull $\text{aff} B$, i.e., the plane containing B , with the cube A itself.

Theorem

Let B be a convex subset of a convex set A in \mathbb{R}^n . Then B is a face of A if and only if $A \setminus B$ is convex and $B = (\text{aff} B) \cap A$. In particular, a point \mathbf{a} of A is an extreme point of A if and only if $A \setminus \{\mathbf{a}\}$ is convex.

- Suppose that B is a face of A . If $\mathbf{x}, \mathbf{y} \in A \setminus B$ and $\lambda, \mu > 0$ with $\lambda + \mu = 1$, then $\lambda \mathbf{x} + \mu \mathbf{y} \in A$, since $\mathbf{x}, \mathbf{y} \in A$ and A is convex. We cannot have $\lambda \mathbf{x} + \mu \mathbf{y} \in B$, for this would imply that $\mathbf{x}, \mathbf{y} \in B$. Thus $\lambda \mathbf{x} + \mu \mathbf{y} \in A \setminus B$ and $A \setminus B$ is convex.

Characterization of the Faces (Cont'd)

- Trivially, $B \subseteq (\text{aff} B) \cap A$.

We now establish the opposite inclusion. Suppose that $\mathbf{u} \in (\text{aff} B) \cap A$. Let $\mathbf{b} \in \text{ri} B$. Then there exist $\mathbf{v} \in B$ and $\alpha, \beta > 0$ with $\alpha + \beta = 1$ such that $\mathbf{b} = \alpha \mathbf{u} + \beta \mathbf{v}$. Since $\mathbf{u}, \mathbf{v} \in A$ and B is a face of A , $\mathbf{u} \in B$. Hence, $(\text{aff} B) \cap A \subseteq B$. So $B = (\text{aff} B) \cap A$.

Suppose next that $A \setminus B$ is convex and $B = (\text{aff} B) \cap A$. If $\lambda \mathbf{x} + \mu \mathbf{y} \in B$, where $\mathbf{x}, \mathbf{y} \in A$ and $\lambda, \mu > 0$ with $\lambda + \mu = 1$, then not both \mathbf{x} and \mathbf{y} can lie in $A \setminus B$, for the convexity of $A \setminus B$ would imply that $\lambda \mathbf{x} + \mu \mathbf{y} \in A \setminus B$. Suppose that $\mathbf{x} \notin A \setminus B$. Then $\mathbf{x} \in B$. So

$$\mathbf{y} = \frac{1}{\mu}(\lambda \mathbf{x} + \mu \mathbf{y}) - \frac{\lambda}{\mu} \mathbf{x} = \left(1 + \frac{\lambda}{\mu}\right)(\lambda \mathbf{x} + \mu \mathbf{y}) - \frac{\lambda}{\mu} \mathbf{x} \in \text{aff} B.$$

Thus, $\mathbf{y} \in (\text{aff} B) \cap A$. Hence, $\mathbf{y} \in B$. So $\mathbf{x}, \mathbf{y} \in B$ and B is a face of A . The final assertion of the theorem follows from what we have just proved and the fact that a singleton set is its own affine hull.

Consequences

Corollary

Each face of a closed convex set in \mathbb{R}^n is closed.

- Let B be a face of a closed convex set A in \mathbb{R}^n . Then B , being the intersection of the closed sets $\text{aff}B$ and A , is itself closed.

Corollary

Let $A = \text{conv}C$, where C is a set in \mathbb{R}^n . Then each extreme point of A lies in C .

- Suppose that there is an extreme point \mathbf{a} of A which does not lie in C . Then $A \setminus \{\mathbf{a}\}$ is a proper convex subset of A containing C . Hence A properly contains $\text{conv}C$, i.e. A . This contradiction shows that each extreme point of A lies in C .

Properties of Faces

Theorem

Let A be a convex set in \mathbb{R}^n . Then:

- (i) The intersection of any non-empty family of faces of A is a face of A ;
- (ii) If B is a face of A , and C is a face of B , then C is a face of A ;
- (iii) The intersection of A with each of its support hyperplanes is a face of A .

- (i) Let $(A_i : i \in I)$ be a non-empty family of faces of A . Then $\bigcap(A_i : i \in I)$ is a convex subset of A . If $\lambda \mathbf{x} + \mu \mathbf{y} \in \bigcap(A_i : i \in I)$, where $\mathbf{x}, \mathbf{y} \in A$ and $\lambda, \mu > 0$ with $\lambda + \mu = 1$, then $\lambda \mathbf{x} + \mu \mathbf{y} \in A_i$ for all $i \in I$. So, since each A_i is a face of A , we have $\mathbf{x}, \mathbf{y} \in A_i$, for all $i \in I$. Hence, $\mathbf{x}, \mathbf{y} \in \bigcap(A_i : i \in I)$. Therefore, $\bigcap(A_i : i \in I)$ is a face of A .

Properties of Faces (Cont'd)

- (ii) Let B be a face of A and let C be a face of B . If $\lambda\mathbf{x} + \mu\mathbf{y} \in C$, where $\mathbf{x}, \mathbf{y} \in A$ and $\lambda, \mu > 0$ with $\lambda + \mu = 1$, then $\lambda\mathbf{x} + \mu\mathbf{y} \in B$, since $C \subseteq B$. Since B is a face of A , $\mathbf{x}, \mathbf{y} \in B$. But C is a face of B , so $\mathbf{x}, \mathbf{y} \in C$. This proves that C is a face of A .
- (iii) Let $H = \{\mathbf{z} \in \mathbb{R}^n : \mathbf{c} \cdot \mathbf{z} = c_0\}$, where $c_0 \in \mathbb{R}$, $\mathbf{c} \in \mathbb{R}^n$ and $\mathbf{c} \neq \mathbf{0}$, be a support hyperplane to A . Suppose that $\mathbf{c} \cdot \mathbf{a} \leq c_0$ whenever $\mathbf{a} \in A$. If $\lambda\mathbf{x} + \mu\mathbf{y} \in A \cap H$, where $\mathbf{x}, \mathbf{y} \in A$ and $\lambda, \mu > 0$ with $\lambda + \mu = 1$, then

$$c_0 = \mathbf{c} \cdot (\lambda\mathbf{x} + \mu\mathbf{y}) = \lambda\mathbf{c} \cdot \mathbf{x} + \mu\mathbf{c} \cdot \mathbf{y} \leq \lambda c_0 + \mu c_0 = c_0.$$

Since $\lambda, \mu > 0$, we must have $\mathbf{c} \cdot \mathbf{x} = \mathbf{c} \cdot \mathbf{y} = c_0$. Hence, $\mathbf{x}, \mathbf{y} \in A \cap H$. Thus, $A \cap H$ is a face of A , for clearly it is a convex subset of A .

Strengthening the Defining Property

Theorem

Let B be a face of a convex set A in \mathbb{R}^n . Suppose that C is a subset of A such that $\text{ri}C$ meets B . Then $C \subseteq B$.

- Let $\mathbf{c} \in C$ and $\mathbf{b} \in B \cap \text{ri}C$. Then there exist $\mathbf{d} \in C$ and $\lambda, \mu > 0$ with $\lambda + \mu = 1$ such that $\mathbf{b} = \lambda\mathbf{c} + \mu\mathbf{d}$. Since $\mathbf{c}, \mathbf{d} \in A$ and B is a face of A , we see that $\mathbf{c} \in B$. Thus, $C \subseteq B$.

Corollary

Let B and C be faces of a convex set A in \mathbb{R}^n such that $\text{ri}B$ and $\text{ri}C$ meet. Then $B = C$.

- Since $\text{ri}C$ meets B , we have $C \subseteq B$. Similarly, $B \subseteq C$. Thus, $B = C$.

On Dimensional Properties of Faces

Corollary

Let B be a face of a convex set A in \mathbb{R}^n , other than A itself. Then $\dim B < \dim A$.

- We suppose that B is non-empty. Clearly, $\text{aff} B \subseteq \text{aff} A$ and $\dim B \leq \dim A$. If $\dim B = \dim A$, then $\text{aff} B = \text{aff} A$ and $\emptyset \subset \text{ri} B \subseteq \text{ri} A$. Hence $A = B$ by the preceding corollary.

Corollary

The intersection of any family of faces of a convex set in \mathbb{R}^n can be expressed as an intersection of $n+1$ or fewer members of the family.

- Suppose that the result is false. Then there exist faces A_1, \dots, A_{n+2} of some convex set A in \mathbb{R}^n such that $A_1 \subset A_2 \subset \dots \subset A_{n+2} \subset A$. Since A_i is a face of A_{i+1} for $i = 1, \dots, n+1$, the preceding corollary shows that

$$-1 < \dim A_1 < \dim A_2 < \dots < \dim A_{n+2} \leq n-1,$$

which is impossible.

Smallest Face Containing a Point

- Each point of a convex set belongs to at least one face of the set, namely the set itself, and in general belongs to several different faces.
Example: A vertex of a three-dimensional cube belongs to one 0-face, three 1-faces, three 2-faces, and one 3-face of the cube.
- Suppose that \mathbf{a} is a point of a convex set A in \mathbb{R}^n and that $F_{\mathbf{a}}$ is the intersection of all faces of A containing \mathbf{a} .
- Then it follows from the preceding theorem that $F_{\mathbf{a}}$ is the **smallest face of A containing \mathbf{a}** .

Characterization of Smallest Face Containing a Point

Theorem

Let \mathbf{a} be a point of a convex set A in \mathbb{R}^n and let $F_{\mathbf{a}}$ be the intersection of all faces of A containing \mathbf{a} . Then $\mathbf{a} \in \text{ri}F_{\mathbf{a}}$ and the relative interiors of the faces of A form a partition of A .

- If $\mathbf{a} \notin \text{ri}F_{\mathbf{a}}$, then $\mathbf{a} \in \text{rebd}F_{\mathbf{a}}$. So, by a previous theorem, there exists a support hyperplane H of $F_{\mathbf{a}}$ passing through \mathbf{a} but not containing $F_{\mathbf{a}}$. Hence, by the preceding theorem, Part (iii), $H \cap F_{\mathbf{a}}$ is a face of A containing \mathbf{a} which is strictly contained in $F_{\mathbf{a}}$. Since this is impossible, $\mathbf{a} \in \text{ri}F_{\mathbf{a}}$. Thus each point \mathbf{a} of A belongs to the relative interior of the face $F_{\mathbf{a}}$ of A . The relative interiors of two different faces of A are disjoint by a previous corollary. Hence the relative interiors of the faces of A form a partition of A .

Smallest Face of a Relative Boundary Point

Corollary

Let \mathbf{a} be a relative boundary point of a convex set A in \mathbb{R}^n . Then $\dim F_{\mathbf{a}} < \dim A$.

- The faces A and $F_{\mathbf{a}}$ of A cannot be equal because $\mathbf{a} \in \text{ri} F_{\mathbf{a}}$. Thus $F_{\mathbf{a}} \subset A$. Hence $\dim F_{\mathbf{a}} < \dim A$ by a previous corollary.

Primitive Convex Sets and Halfflats

- A closed convex set in \mathbb{R}^n which is not the convex hull of its relative boundary is said to be **primitive**.
- The reader should have little difficulty in discovering that the only primitive sets in \mathbb{R}^2 are:
 - points;
 - lines;
 - halflines;
 - closed halfplanes;
 - \mathbb{R}^2 itself.
- Before we can extend this last result to \mathbb{R}^n , we need to generalize the concepts of **halflines** and **halfplanes** from \mathbb{R}^2 to \mathbb{R}^n .
- In \mathbb{R}^n a **closed halfflat** is the intersection of a flat with a closed halfspace which meets it, but does not contain it.

Characterization of Primitive Sets

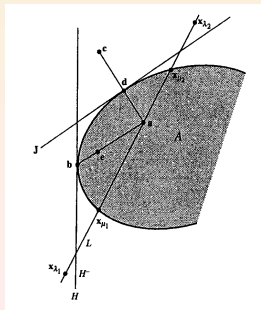
Theorem

A closed convex set in \mathbb{R}^n is primitive if and only if it is either a nonempty flat or a closed halfflat.

- We establish only the non-trivial part of the theorem, i.e., that, for each closed convex set A in \mathbb{R}^n other than a flat or a closed halfflat, $A = \text{conv}(\text{rebd}A)$.

We know $\text{conv}(\text{rebd}A) \subseteq A$. Moreover, $A = (\text{rebd}A) \cup (\text{ri}A)$ and $\text{rebd}A \subseteq \text{conv}(\text{rebd}A)$. So, we must show $\text{ri}A \subseteq \text{conv}(\text{rebd}A)$.

Let $\mathbf{a} \in \text{ri}A$. Since A is not a flat, its relative boundary is not empty, say $\mathbf{b} \in \text{rebd}A$. A previous theorem shows that there is a non-trivial support hyperplane H to A at \mathbf{b} .



Characterization of Primitive Sets (Cont'd)

- Since $A \cap H$ is a proper face of A , a previous theorem shows that $\mathbf{a} \notin H$. Thus there exist $u_0 \in \mathbb{R}$, $\mathbf{u} \in \mathbb{R}^n$ such that H has equation $\mathbf{n} \cdot \mathbf{x} = u_0$ and $\mathbf{u} \cdot \mathbf{a} < u_0$, $\mathbf{u} \cdot \mathbf{b} = u_0$ with A lying in the closed halfspace

$$H^- = \{\mathbf{x} \in \mathbb{R}^n : \mathbf{u} \cdot \mathbf{x} \leq u_0\}.$$

By the hypothesis, A is not a closed halfflat, and so cannot be $(\text{aff}A) \cap H^-$. Thus there exists a point \mathbf{c} of $(\text{aff}A) \cap H^-$ that does not lie in A . Denote by \mathbf{d} the point where the line segment joining \mathbf{a} and \mathbf{c} meets $\text{rebd}A$. Since $\mathbf{u} \cdot \mathbf{a} < u_0$ and $\mathbf{u} \cdot \mathbf{c} \leq u_0$, we get $\mathbf{u} \cdot \mathbf{d} < u_0$.

The existence of a non-trivial support hyperplane to A at \mathbf{d} , J say, shows that there exist $v_0 \in \mathbb{R}$, $\mathbf{v} \in \mathbb{R}^n$ such that J has equation $\mathbf{v} \cdot \mathbf{x} = v_0$ and $\mathbf{v} \cdot \mathbf{a} < v_0$, $\mathbf{v} \cdot \mathbf{d} = v_0$, with A lying in the closed halfspace $\{\mathbf{x} \in \mathbb{R}^n : \mathbf{v} \cdot \mathbf{x} \leq v_0\}$.

Characterization of Primitive Sets (Cont'd)

- Since $\mathbf{u} \cdot \mathbf{a} < u_0$, $\mathbf{u} \cdot \mathbf{b} = u_0$ and $\mathbf{u} \cdot \mathbf{d} < u_0$, there is a point \mathbf{e} of $\text{ri}A$ lying on the line segment joining \mathbf{a} and \mathbf{b} such that $\mathbf{u} \cdot \mathbf{e} > \mathbf{u} \cdot \mathbf{d}$. Because \mathbf{e} lies in $\text{ri}A$, we must have $\mathbf{v} \cdot \mathbf{e} < v_0$. For each scalar λ , denote by \mathbf{x}_λ the point $\mathbf{a} + \lambda(\mathbf{d} - \mathbf{e})$ on the line L joining the points \mathbf{a} and $\mathbf{a} + \mathbf{d} - \mathbf{e}$ of $\text{aff}A$. Choose scalars λ_1, λ_2 such that

$$\lambda_1 < \frac{u_0 - \mathbf{u} \cdot \mathbf{a}}{\mathbf{u} \cdot \mathbf{d} - \mathbf{u} \cdot \mathbf{e}} < 0 < \frac{v_0 - \mathbf{v} \cdot \mathbf{a}}{\mathbf{v} \cdot \mathbf{d} - \mathbf{v} \cdot \mathbf{e}} < \lambda_2.$$

Then

$$\begin{aligned} \mathbf{u} \cdot \mathbf{x}_{\lambda_1} &= \mathbf{u} \cdot (\mathbf{a} + \lambda_1(\mathbf{d} - \mathbf{e})) = \mathbf{u} \cdot \mathbf{a} + \lambda_1 \mathbf{u} \cdot (\mathbf{d} - \mathbf{e}) \\ &> \mathbf{u} \cdot \mathbf{a} + u_0 - \mathbf{u} \cdot \mathbf{a} = u_0; \\ \mathbf{v} \cdot \mathbf{x}_{\lambda_2} &= \mathbf{v} \cdot (\mathbf{a} + \lambda_2(\mathbf{d} - \mathbf{e})) = \mathbf{v} \cdot \mathbf{a} + \lambda_2 \mathbf{v} \cdot (\mathbf{d} - \mathbf{e}) \\ &> \mathbf{v} \cdot \mathbf{a} + v_0 - \mathbf{v} \cdot \mathbf{a} = v_0. \end{aligned}$$

Hence neither \mathbf{x}_{λ_1} nor \mathbf{x}_{λ_2} lies in A . Thus, there are scalars μ_1, μ_2 with $\lambda_1 < \mu_1 < 0 < \mu_2 < \lambda_2$ such that $\mathbf{x}_{\mu_1}, \mathbf{x}_{\mu_2} \in \text{rebd}A$. Hence, $\mathbf{a} = \frac{\mu_2}{\mu_2 - \mu_1} \mathbf{x}_{\mu_1} - \frac{\mu_1}{\mu_2 - \mu_1} \mathbf{x}_{\mu_2} \in \text{conv}\{\mathbf{x}_{\mu_1}, \mathbf{x}_{\mu_2}\}$. So $\text{ri}A \subseteq \text{conv}(\text{rebd}A)$.

Facial Structure of Closed Convex Sets

- By the convex hull of a family of sets in \mathbb{R}^n is meant the convex hull of its union.

Theorem

Every closed convex set in \mathbb{R}^n is the convex hull of its primitive faces.

- Let A be a closed convex set in \mathbb{R}^n . We argue by induction on the dimension of A .

The case $\dim A = -1$ is trivial.

Suppose that $\dim A = m$, where $m > -1$, and that the assertion is true for all closed convex sets in \mathbb{R}^n with dimension less than m . The theorem is trivial when A is primitive. Suppose, then, that A is not primitive. Denote by B the convex hull of the primitive faces of A . Then $B \subseteq A$. So we need only show that $A \subseteq B$.

Facial Structure of Closed Convex Sets (Cont'd)

- Since A is not primitive, we have $A = \text{conv}(\text{rebd}A)$. Let $\mathbf{a} \in \text{rebd}A$. Then \mathbf{a} lies in $F_{\mathbf{a}}$, the smallest face of A containing \mathbf{a} . By a previous corollary, $\dim F_{\mathbf{a}} < \dim A$. The induction hypothesis shows that $F_{\mathbf{a}}$ is the convex hull of its primitive faces. Since each primitive face of $F_{\mathbf{a}}$ is a primitive face of A , $F_{\mathbf{a}} \subseteq B$. Hence, $\mathbf{a} \in F_{\mathbf{a}} \subseteq B$ and $\text{rebd}A \subseteq B$. So $A = \text{conv}(\text{rebd}A) \subseteq B$. Thus, $A = B$, i.e., A is the convex hull of its primitive faces.

Consequences

Corollary

Every closed convex set in \mathbb{R}^n is the convex hull of those of its faces which are flats or closed halfflats.

- The result follows from the theorem and a previous theorem.

Corollary

Every closed convex set in \mathbb{R}^n that contains no lines is the convex hull of its extreme points and extreme halflines.

- The corollary follows from the theorem and the fact that points and halflines are the only primitive sets which contain no lines.

Theorem (Krein-Milman)

Every compact convex set in \mathbb{R}^n is the convex hull of its extreme points.

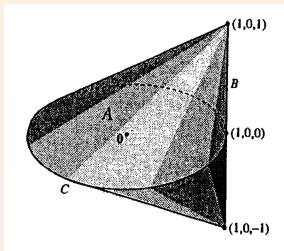
- The theorem follows from the preceding theorem and the fact that points are the only compact primitive sets.

Comment on Set of Extreme Points

- The Krein-Milman theorem shows that the convex hull of the extreme points of a compact convex set in \mathbb{R}^n is closed.
- It is not true, however, that the set of extreme points itself is necessarily closed.
- To see this, let A and B denote the circular disc and the line segment in \mathbb{R}^3 given by the equations

$$A = \{(x, y, 0) : x^2 + y^2 \leq 1\} \quad \text{and} \quad B = \{(1, 0, z) : -1 \leq z \leq 1\}.$$

Let $C = \text{conv}(A \cup B)$. Then C is a compact convex set. Its set of extreme points consists of $(1, 0, 1)$ and $(1, 0, -1)$ together with the points on the relative boundary of A with the exception of $(1, 0, 0)$. This set is not closed.



Exposed Faces

- By a previous theorem, the intersection of a convex set in \mathbb{R}^n with one of its support hyperplanes is a face of the set.
- A face which arises in this way is called an **exposed face** of the set.
- It is technically convenient to allow the empty set and the set itself as exposed faces of any convex set in \mathbb{R}^n .
- Thus, an **exposed face** of a convex set in \mathbb{R}^n is either the empty set, the set itself, or the intersection of the set with one of its support hyperplanes.

Example: Exposed Faces

- Faces of a convex set are not always exposed.

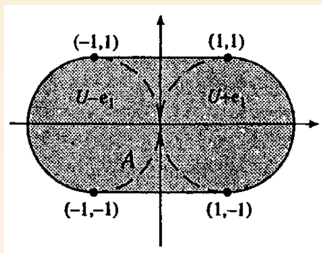
Example: Let A be the convex hull of the union of the circular discs $U + \mathbf{e}_1$, and $U - \mathbf{e}_1$, in \mathbb{R}^2 , where

$$U = \{(x, y) : x^2 + y^2 \leq 1\} \quad \text{and} \quad \mathbf{e}_1 = (1, 0).$$

Then the points $(1, 1)$, $(1, -1)$, $(-1, 1)$, $(-1, -1)$ are faces of A that are not exposed.

- This example also serves to show that an exposed face of an exposed face of a convex set need not be an exposed face of that convex set.

The line segment joining $(-1, 1)$ and $(1, 1)$ is an exposed face of A and $(1, 1)$ is an exposed face of this line segment, but $(1, 1)$ is not an exposed face of A .



Closure of Exposed Faces Under Intersections

Theorem

The intersection of any non-empty family of exposed faces of a convex set in \mathbb{R}^n is an exposed face of the set.

- In view of a previous corollary, we may assume that the family of exposed faces is finite.

Let A_1, \dots, A_m be exposed faces of a convex set A in \mathbb{R}^n . We show that $A_1 \cap \dots \cap A_m$ is an exposed face of A , considering only the non-trivial case when the A_1, \dots, A_m are proper exposed faces of A , whose intersection is non-empty, containing some point \mathbf{a}_0 , say. For each $i = 1, \dots, m$, there exists $\mathbf{u}_i \in \mathbb{R}^n$ such that

$$A_i = \{\mathbf{a} \in A : \mathbf{u}_i \cdot \mathbf{a} = \mathbf{u}_i \cdot \mathbf{a}_0\} \quad \text{and} \quad A \subseteq \{\mathbf{x} \in \mathbb{R}^n : \mathbf{u}_i \cdot \mathbf{x} \leq \mathbf{u}_i \cdot \mathbf{a}_0\}.$$

Closure of Exposed Faces Under Intersections (Cont'd)

- It follows easily that

$$A_1 \cap \cdots \cap A_m = \{\mathbf{a} \in A : (\mathbf{u}_1 + \cdots + \mathbf{u}_m) \cdot \mathbf{a} = (\mathbf{u}_1 + \cdots + \mathbf{u}_m) \cdot \mathbf{a}_0\},$$

and that

$$A \subseteq \{\mathbf{x} \in \mathbb{R}^n : (\mathbf{u}_1 + \cdots + \mathbf{u}_m) \cdot \mathbf{x} \leq (\mathbf{u}_1 + \cdots + \mathbf{u}_m) \cdot \mathbf{a}_0\}.$$

This shows that $A_1 \cap \cdots \cap A_m$ is an exposed face of A .

Exposed Faces of Sums

Theorem

Let A and B be convex sets in \mathbb{R}^n . Then each exposed face of $A+B$ has the form $C+D$, where C is an exposed face of A and D is an exposed face of B .

- Suppose that F is a *proper* exposed face of $A+B$. Then there exist $\mathbf{a}_0 \in A$, $\mathbf{b}_0 \in B$, and a non-zero \mathbf{u} in \mathbb{R}^n such that

$$\begin{aligned} A+B &\subseteq \{\mathbf{x} \in \mathbb{R}^n : \mathbf{u} \cdot \mathbf{x} \leq \mathbf{u} \cdot (\mathbf{a}_0 + \mathbf{b}_0)\}, \\ F &= \{\mathbf{x} \in A+B : \mathbf{u} \cdot \mathbf{x} = \mathbf{u} \cdot (\mathbf{a}_0 + \mathbf{b}_0)\}. \end{aligned}$$

If $\mathbf{a} \in A$, then $\mathbf{a} + \mathbf{b}_0 \in A+B$. Hence $\mathbf{u} \cdot (\mathbf{a} + \mathbf{b}_0) \leq \mathbf{u} \cdot (\mathbf{a}_0 + \mathbf{b}_0)$ and $\mathbf{u} \cdot \mathbf{a} \leq \mathbf{u} \cdot \mathbf{a}_0$. Similarly, if $\mathbf{b} \in B$, then $\mathbf{u} \cdot \mathbf{b} \leq \mathbf{u} \cdot \mathbf{b}_0$. Thus

$$C = \{\mathbf{x} \in A : \mathbf{u} \cdot \mathbf{x} = \mathbf{u} \cdot \mathbf{a}_0\} \quad \text{and} \quad D = \{\mathbf{x} \in B : \mathbf{u} \cdot \mathbf{x} = \mathbf{u} \cdot \mathbf{b}_0\}$$

are, respectively, exposed faces of A and B .

Exposed Faces of Sums (Cont'd)

- We derived that

$$C = \{\mathbf{x} \in A : \mathbf{u} \cdot \mathbf{x} = \mathbf{u} \cdot \mathbf{a}_0\} \quad \text{and} \quad D = \{\mathbf{x} \in B : \mathbf{u} \cdot \mathbf{x} = \mathbf{u} \cdot \mathbf{b}_0\}$$

are, respectively, exposed faces of A and B . Clearly $C + D \subseteq F$.

If $\mathbf{f} \in F$, then $\mathbf{f} = \mathbf{a} + \mathbf{b}$ for some $\mathbf{a} \in A$, $\mathbf{b} \in B$.

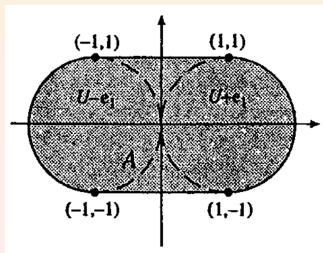
Now $\mathbf{u} \cdot (\mathbf{a} + \mathbf{b}) = \mathbf{u} \cdot (\mathbf{a}_0 + \mathbf{b}_0)$, $\mathbf{u} \cdot \mathbf{a} \leq \mathbf{u} \cdot \mathbf{a}_0$ and $\mathbf{u} \cdot \mathbf{b} \leq \mathbf{u} \cdot \mathbf{b}_0$.

Hence, $\mathbf{u} \cdot \mathbf{a} = \mathbf{u} \cdot \mathbf{a}_0$ and $\mathbf{u} \cdot \mathbf{b} = \mathbf{u} \cdot \mathbf{b}_0$.

Thus $\mathbf{a} \in C$, $\mathbf{b} \in D$, and $F = C + D$.

Exposed Points

- The zero-dimensional exposed faces of a convex set are called its **exposed points**.
- Thus a point \mathbf{a} of a convex set A in \mathbb{R}^n is an exposed point of A if and only if there is some support hyperplane to A meeting it in the single point \mathbf{a} .
- Every exposed point of a convex set in \mathbb{R}^n is one of its extreme points, but not necessarily conversely.
- The point $(1,1)$ of the set A of the preceding example is an extreme, but not an exposed, point of A .



Farthest Points

- Let A be a non-empty compact set in \mathbb{R}^n and let \mathbf{b} be a point of \mathbb{R}^n .
- For each point \mathbf{x} of A , denote by $f(\mathbf{x})$ the distance $\|\mathbf{x} - \mathbf{b}\|$ of \mathbf{x} from \mathbf{b} .
- Then f is a continuous real-valued function defined on a non-empty compact set A .
- So it is bounded and attains its bounds.
- In particular, f attains its upper bound.
- So there is a point \mathbf{a} of A such that $\|\mathbf{x} - \mathbf{b}\| \leq \|\mathbf{a} - \mathbf{b}\|$ for all \mathbf{x} in A .
- Each such point \mathbf{a} of A is called a **farthest point** of A from \mathbf{b} .

Farthest Points and Exposed Points

Theorem

Let \mathbf{a} be a farthest point of a compact convex set A in \mathbb{R}^n from some point \mathbf{b} of \mathbb{R}^n . Then \mathbf{a} is an exposed point of A .

- We consider the non-trivial case when $\mathbf{a} \neq \mathbf{b}$. Since \mathbf{a} is a farthest point of A from \mathbf{b} , we have, for each point \mathbf{x} of A ,

$$\begin{aligned} \|\mathbf{a} - \mathbf{b}\|^2 &\geq \|\mathbf{x} - \mathbf{b}\|^2 \\ &= \|(\mathbf{x} - \mathbf{a}) + (\mathbf{a} - \mathbf{b})\|^2 \\ &= \|\mathbf{x} - \mathbf{a}\|^2 + 2(\mathbf{x} - \mathbf{a}) \cdot (\mathbf{a} - \mathbf{b}) + \|\mathbf{a} - \mathbf{b}\|^2. \end{aligned}$$

Hence $0 \geq (\mathbf{x} - \mathbf{a}) \cdot (\mathbf{a} - \mathbf{b})$. Equality occurs in the last inequality if and only if $\mathbf{x} = \mathbf{a}$. So the hyperplane $H = \{\mathbf{z} \in \mathbb{R}^n : (\mathbf{a} - \mathbf{b}) \cdot (\mathbf{z} - \mathbf{a}) = 0\}$ supports A at \mathbf{a} and $H \cap A = \{\mathbf{a}\}$. Thus \mathbf{a} is an exposed point of A .

Compact Convex Sets and Exposed Points

Lemma

A compact convex set in \mathbb{R}^n has an exposed point in every open halfspace which meets it.

- Suppose that the compact convex set A in \mathbb{R}^n meets the open halfspace $J = \{\mathbf{z} \in \mathbb{R}^n : \mathbf{u} \cdot \mathbf{z} + u_0 < 0\}$, where $u_0 \in \mathbb{R}$, $\mathbf{u} \in \mathbb{R}^n$, and $\mathbf{u} \neq \mathbf{0}$, say $\mathbf{a} \in A \cap J$. Let $\lambda > 0$ satisfy $s^2 + 2\lambda(\mathbf{u} \cdot \mathbf{a} + u_0) < 0$, where s is the diameter of A . Let $\mathbf{c} = \mathbf{a} + \lambda\mathbf{u}$. For each point \mathbf{x} of $A \setminus J$, $\mathbf{u} \cdot \mathbf{x} + u_0 \geq 0$ and $\mathbf{u} \cdot (\mathbf{a} - \mathbf{x}) \leq \mathbf{u} \cdot \mathbf{a} + u_0$. Hence,

$$\begin{aligned} \|\mathbf{c} - \mathbf{x}\|^2 &= \|\mathbf{a} - \mathbf{x} + \lambda\mathbf{u}\|^2 \\ &= \|\mathbf{a} - \mathbf{x}\|^2 + 2\lambda\mathbf{u} \cdot (\mathbf{a} - \mathbf{x}) + \lambda^2\|\mathbf{u}\|^2 \\ &\leq s^2 + 2\lambda(\mathbf{u} \cdot \mathbf{a} + u_0) + \lambda^2\|\mathbf{u}\|^2 \\ &< \lambda^2\|\mathbf{u}\|^2 = \|\mathbf{c} - \mathbf{a}\|^2. \end{aligned}$$

Thus no point \mathbf{x} of $A \setminus J$ is a farthest point of A from \mathbf{c} . So every farthest point of A from \mathbf{c} is an exposed point of A lying in J .

Compact Convex Sets as Hulls of Exposed Points

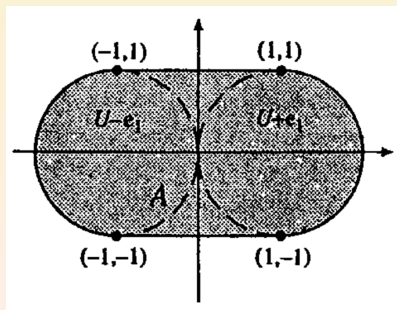
Theorem (Straszewicz)

Every compact convex set in \mathbb{R}^n is the closure of the convex hull of its exposed points.

- Let B be the set of the exposed points of a compact convex set A in \mathbb{R}^n . Trivially $\text{cl}(\text{conv}B) \subseteq A$. So we must show that $A \subseteq \text{cl}(\text{conv}B)$. Suppose that this is not so. Then there is a point \mathbf{a} of A which does not belong to the closed convex set $\text{cl}(\text{conv}B)$. It follows immediately from a previous corollary that there is an open halfspace J in \mathbb{R}^n which contains \mathbf{a} but is disjoint from $\text{cl}(\text{conv}B)$. By the preceding lemma, there is a point of B lying in J , which is impossible. Thus $A \subseteq \text{cl}(\text{conv}B)$ as desired.

Necessity of the Closure Requirement

- The two-dimensional set illustrated in the figure



shows that the closure requirement in Straszewicz's theorem cannot be omitted.

Subsection 7

The Blaschke Selection Principle

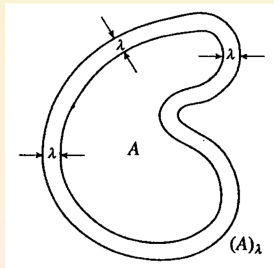
Distance Between Two Sets

- Sets A and B in \mathbb{R}^n are said to be a **finite distance apart** if there exists $\lambda \geq 0$ such that, for each point \mathbf{a} of A , there is a point \mathbf{b} of B whose distance $\|\mathbf{a} - \mathbf{b}\|$ from \mathbf{a} does not exceed λ , and vice versa.
- In this situation, we say that the **distance between A and B does not exceed λ** .
- The **distance between** sets A and B in \mathbb{R}^n that are a finite distance apart is defined to be the infimum of the set of all those $\lambda \geq 0$ for which the distance between A and B does not exceed λ .
- This definition does not assign a distance between the empty set and a non-empty set or between a bounded set and an unbounded one.
- On the other hand, the definition always assigns a distance between two non-empty bounded sets.
- For our purposes here, it will be sufficient to restrict attention to non-empty compact sets.

λ -Neighborhood of a Set

- Let A be a set in \mathbb{R}^n and let $\lambda \geq 0$.
- Then the λ -neighborhood $(A)_\lambda$ of A is the set $A + \lambda U$, where U denotes the closed unit ball $\{\mathbf{x} \in \mathbb{R}^n : \|\mathbf{x}\| \leq 1\}$.

The figure makes it clear why the set $(A)_\lambda$ is often referred to as the **outer parallel set of A at distance λ** .



- Clearly, if $\mathbf{a} \in \mathbb{R}^n$, $r > 0$, and $\lambda \geq 0$, then:
 - The r -neighborhood of $\{\mathbf{a}\}$ is the closed ball $B[\mathbf{a}; r]$;
 - The λ -neighborhood of $B[\mathbf{a}; r]$ is the closed ball $B[\mathbf{a}; r + \lambda]$.

Properties of Neighborhoods

Theorem

Let A, B be sets in \mathbb{R}^n and let $\lambda, \mu \geq 0$. Then:

- (i) $(A)_0 = A$ and $A \subseteq (A)_\lambda$;
- (ii) $(A)_\lambda \subseteq (B)_\lambda$ when $A \subseteq B$;
- (iii) $(A)_\lambda$ is convex when A is;
- (iv) $((A)_\lambda)_\mu = (A)_{\lambda+\mu}$.

- Parts (i) and (ii) are easy consequences of the definition of a λ -neighborhood.

Part (iii) follows from a previous example and theorem.

To prove Part (iv) we note, using a previous theorem, that

$$((A)_\lambda)_\mu = (A)_\lambda + \mu U = (A + \lambda U) + \mu U = A + (\lambda + \mu)U = (A)_{\lambda+\mu}.$$

The Hausdorff Distance

- The assertions that, for each point \mathbf{a} of a set A in \mathbb{R}^n , there is a point \mathbf{b} of a set B in \mathbb{R}^n such that $\|\mathbf{a} - \mathbf{b}\| \leq \lambda$, and that $A \subseteq (B)_\lambda$, where $\lambda \geq 0$, are equivalent.
- Thus the definition we now give of the distance between non-empty compact sets in \mathbb{R}^n coincides with the one given earlier.
- The **distance** $\rho(A, B)$ between non-empty compact sets A, B in \mathbb{R}^n is defined by the equation

$$\rho(A, B) = \inf \{ \lambda \geq 0 : A \subseteq (B)_\lambda \text{ and } B \subseteq (A)_\lambda \}.$$

- The assumptions that A and B are non-empty and compact ensure that $\rho(A, B)$ is well-defined.
- The function ρ is known as the **Hausdorff metric** or **Hausdorff distance**.

Properties of the Distance

- It is easily seen that the Hausdorff distance $\rho(\{\mathbf{a}\}, \{\mathbf{b}\})$ between the singleton sets $\{\mathbf{a}\}$ and $\{\mathbf{b}\}$ in \mathbb{R}^n is $\|\mathbf{a} - \mathbf{b}\|$, i.e., the distance between the points \mathbf{a} and \mathbf{b} themselves.
- Another readily verified fact is that the Hausdorff distance is invariant under translation in the sense that, if A and B are non-empty compact sets in \mathbb{R}^n and \mathbf{x} is a point of \mathbb{R}^n , then $\rho(A, B) = \rho(A + \mathbf{x}, B + \mathbf{x})$.

Example

- Let A and B be, respectively, the closed balls $B[\mathbf{a}; r]$ and $B[\mathbf{b}; s]$ in \mathbb{R}^n . Then $\rho(A, B) = \|\mathbf{b} - \mathbf{a}\| + |s - r|$.
- Suppose first that $r \leq s$. We have

$$\begin{aligned} A &\subseteq B - (\mathbf{b} - \mathbf{a}) \subseteq (B)_{\|\mathbf{b} - \mathbf{a}\|}; \\ B &= A + \mathbf{b} - \mathbf{a} + (s - r)U \subseteq (A)_{\|\mathbf{b} - \mathbf{a}\| + s - r}. \end{aligned}$$

Hence, $\rho(A, B) \leq \|\mathbf{b} - \mathbf{a}\| + s - r$.

Now B contains a point whose distance from \mathbf{a} is $\|\mathbf{b} - \mathbf{a}\| + s$.

Thus, if $\lambda \geq 0$ and $B \subseteq (A)_\lambda = B[\mathbf{a}; \lambda + r]$, then $\|\mathbf{b} - \mathbf{a}\| + s \leq \lambda + r$.

Hence, $\rho(A, B) \geq \|\mathbf{b} - \mathbf{a}\| + s - r$. Thus, $\rho(A, B) = \|\mathbf{b} - \mathbf{a}\| + s - r$.

I.e., $\rho(A, B) = \|\mathbf{b} - \mathbf{a}\| + |s - r|$.

The case $s \leq r$ is similar.

Necessary Condition

Theorem

Let A and B be nonempty compact sets in \mathbb{R}^n with $\rho(A, B) = \lambda$. Then $A \subseteq (B)_\lambda$ and $B \subseteq (A)_\lambda$.

- Let $\mathbf{a} \in A$. For each $\varepsilon > 0$, $A \subseteq (B)_{\lambda+\varepsilon}$. Hence there is a point \mathbf{b}_ε of B for which $\|\mathbf{a} - \mathbf{b}_\varepsilon\| \leq \lambda + \varepsilon$. So $\inf \{\|\mathbf{a} - \mathbf{b}\| : \mathbf{b} \in B\} \leq \lambda$.

A previous theorem shows that there exists some point \mathbf{b}_0 of B such that $\|\mathbf{a} - \mathbf{b}_0\| \leq \lambda$. Thus, $\mathbf{a} \in (B)_\lambda$ and $A \subseteq (B)_\lambda$.

Similarly, $B \subseteq (A)_\lambda$.

Metric Properties of the Distance

Theorem

Let A, B, C be non-empty compact sets in \mathbb{R}^n and let $\theta \geq 0$. Then:

- (i) $\rho(A, B) \geq 0$ and $\rho(A, B) = 0$ if and only if $A = B$;
 - (ii) $\rho(A, B) = \rho(B, A)$;
 - (iii) $\rho(A, C) \leq \rho(A, B) + \rho(B, C)$;
 - (iv) $\rho(\text{conv}A, \text{conv}B) \leq \rho(A, B)$;
 - (v) if A and B are convex, then $\rho(A, B) = \rho((A)_\theta, (B)_\theta)$.
- (i) Trivially $\rho(A, B) \geq 0$. Also $\rho(A, A) = 0$. If $\rho(A, B) = 0$, then $A \subseteq (B)_0 = B$ and $B \subseteq (A)_0 = A$. Hence $A = B$.
- (ii) This follows immediately from the definition of ρ .
- (iii) Let $\rho(A, B) = \alpha$ and $\rho(B, C) = \beta$. Then $A \subseteq (B)_\alpha \subseteq ((C)_\beta)_\alpha = (C)_{\alpha+\beta}$ and $C \subseteq (B)_\beta \subseteq ((A)_\alpha)_\beta = (A)_{\alpha+\beta}$. Hence $\rho(A, C) \leq \alpha + \beta = \rho(A, B) + \rho(B, C)$.

Metric Properties of the Distance (Cont'd)

- (iv) Let $\rho(A, B) = \alpha$. Then $(\text{conv}A)_\alpha$ is convex and $B \subseteq (A)_\alpha \subseteq (\text{conv}A)_\alpha$. Hence $\text{conv}B \subseteq (\text{conv}A)_\alpha$. Similarly, $\text{conv}A \subseteq (\text{conv}B)_\alpha$. Thus, $\rho(\text{conv}A, \text{conv}B) \leq \alpha = \rho(A, B)$. We note that $\text{conv}A$ and $\text{conv}B$ are compact by a previous theorem.
- (v) Let A and B be convex. The sets $(A)_\theta$ and $(B)_\theta$ are compact by a previous theorem. Let $\rho(A, B) = \alpha$ and $\rho((A)_\theta, (B)_\theta) = \beta$. Then

$$(A)_\theta \subseteq ((B)_\alpha)_\theta = ((B)_\theta)_\alpha \quad \text{and} \quad (B)_\theta \subseteq ((A)_\alpha)_\theta = ((A)_\theta)_\alpha.$$

This shows that $\beta \leq \alpha$. Also

$$A + \theta U \subseteq (B + \theta U) + \beta U \quad \text{and} \quad B + \theta U \subseteq (A + \theta U) + \beta U,$$

i.e.,

$$A + \theta U \subseteq (B + \beta U) + \theta U \quad \text{and} \quad B + \theta U \subseteq (A + \beta U) + \theta U.$$

Hence, $A \subseteq B + \beta U$ and $B \subseteq A + \beta U$ by a previous theorem. Thus, $\alpha \leq \beta$. So $\alpha = \beta$.

Convergence of Compact Sets

- The sequence A_1, \dots, A_j, \dots of non-empty compact sets in \mathbb{R}^n is said to **converge** to the non-empty compact set A in \mathbb{R}^n , written $A_j \rightarrow A$ as $j \rightarrow \infty$, if $\rho(A_j, A) \rightarrow 0$ as $j \rightarrow \infty$.
- Such a sequence cannot converge to more than one nonempty compact set A .

If it also converges to a nonempty compact set B in \mathbb{R}^n , then

$$0 \leq \rho(A, B) \leq \rho(A, A_j) + \rho(A_j, B) \rightarrow 0,$$

as $j \rightarrow \infty$. Hence $\rho(A, B) = 0$ and $A = B$.

Convergence and Convexity

- If the sequence A_1, \dots, A_j, \dots converges to A and each A_j is convex, then so too is A .

Suppose that a sequence A_1, \dots, A_j, \dots of nonempty compact convex sets in \mathbb{R}^n converges to a nonempty compact set A in \mathbb{R}^n . The preceding theorem shows that

$$\rho(A_j, \text{conv}A) = \rho(\text{conv}A_j, \text{conv}A) \leq \rho(A_j, A) \rightarrow 0$$

as $j \rightarrow \infty$. Thus A_1, \dots, A_j, \dots also converges to $\text{conv}A$. Since a sequence cannot converge to two different limits, $A = \text{conv}A$ and A is convex.

Convergence and Linear Combinations

Theorem

Let sequences A_1, \dots, A_j, \dots and B_1, \dots, B_j, \dots converge, respectively, to A_0 and B_0 , where all the A 's and B 's are nonempty compact sets in \mathbb{R}^n . Let real sequences $\alpha_1, \dots, \alpha_j, \dots$ and $\beta_1, \dots, \beta_j, \dots$ converge, respectively, to α and β . Then the sequence $\alpha_1 A_1 + \beta_1 B_1, \dots, \alpha_j A_j + \beta_j B_j, \dots$ converges to $\alpha A_0 + \beta B_0$.

- For $\mathbf{a}, \mathbf{b}, \mathbf{a}_j, \mathbf{b}_j \in \mathbb{R}^n$,

$$\begin{aligned} & \|\alpha_j \mathbf{a}_j + \beta_j \mathbf{b}_j - \alpha \mathbf{a} - \beta \mathbf{b}\| \\ &= \|\alpha_j(\mathbf{a}_j - \mathbf{a}) + (\alpha_j - \alpha)\mathbf{a} + \beta_j(\mathbf{b}_j - \mathbf{b}) + (\beta_j - \beta)\mathbf{b}\| \\ &\leq |\alpha_j| \|\mathbf{a}_j - \mathbf{a}\| + |\alpha_j - \alpha| \|\mathbf{a}\| + |\beta_j| \|\mathbf{b}_j - \mathbf{b}\| + |\beta_j - \beta| \|\mathbf{b}\|. \end{aligned}$$

Write $\theta_j = \rho(A_j, A_0)$ and $\varphi_j = \rho(B_j, B_0)$. Let $r > 0$ be such that $\|\mathbf{a}\|, \|\mathbf{b}\| < r$ whenever $\mathbf{a} \in A_0, \mathbf{b} \in B_0$.

Convergence and Linear Combinations (Cont'd)

- It follows easily from the above inequality that

$$\begin{aligned} \rho(\alpha_j A_j + \beta_j B_j, \alpha A_0 + \beta B_0) \\ \leq |\alpha_j| \theta_j + |\alpha_j - \alpha| r + |\beta_j| \varphi_j + |\beta_j - \beta| r. \end{aligned}$$

Hence, $\alpha_j A_j + \beta_j B_j \rightarrow \alpha A_0 + \beta B_0$.

Corollary

For $i = 1, \dots, m$, let the sequence $A_1^i, \dots, A_j^i, \dots$ converge to A_0^i , where all the A 's are nonempty compact sets in \mathbb{R}^n . Let the real sequence $\alpha_1^i, \dots, \alpha_j^i, \dots$ converge to α_i . Then $\alpha_j^1 A_j^1 + \dots + \alpha_j^m A_j^m \rightarrow \alpha_1 A_0^1 + \dots + \alpha_m A_0^m$ as $j \rightarrow \infty$.

Convex Bodies, Inradius and Circumradius

- A **convex body** C is a compact convex set in \mathbb{R}^n that has a nonempty interior.
- The **inradius** r of C is the supremum of the set of radii of closed balls lying in C .
- The **circumradius** R of C is the infimum of the set of radii of closed balls in \mathbb{R}^n containing C .
- Clearly both r and R are positive real numbers satisfying $r \leq R$.

Convex Bodies and Balls

Theorem

Let C be a convex body in \mathbb{R}^n with inradius r and circumradius R . Then C contains a closed ball of radius r and is contained in a unique closed ball of radius R .

- The definition of R implies that, for each, $j = 1, 2, \dots$, there exist $\mathbf{a}_j \in \mathbb{R}^n$ and $R_j > 0$ such that $C \subseteq B[\mathbf{a}_j; R_j]$ and $R_j < R + \frac{1}{j}$. The sequence R_1, \dots, R_j, \dots converges to R . The sequence $\mathbf{a}_1, \dots, \mathbf{a}_j, \dots$ is bounded. Thus there is some subsequence $\mathbf{a}_{i_1}, \dots, \mathbf{a}_{i_j}, \dots$ of $\mathbf{a}_1, \dots, \mathbf{a}_j, \dots$ that converges to some point \mathbf{a} of \mathbb{R}^n . It follows from a previous example that $B[\mathbf{a}_{i_j}; R_{i_j}] \rightarrow B[\mathbf{a}; R]$ as $j \rightarrow \infty$.

We show that $C \subseteq B[\mathbf{a}; R]$: Let $\mathbf{c} \in C$. Since $C \subseteq B[\mathbf{a}_{i_j}; R_{i_j}]$, $\|\mathbf{c} - \mathbf{a}_{i_j}\| \leq R_{i_j}$. Letting $j \rightarrow \infty$ in the last inequality, we find that $\|\mathbf{c} - \mathbf{a}\| \leq R$. Thus, $\mathbf{c} \in B[\mathbf{a}; R]$. So $C \subseteq B[\mathbf{a}; R]$.

Convex Bodies and Balls (Cont'd)

- The proof that C contains a closed ball of radius r is similar to the one which we have just given.

Suppose that C lies in both of the closed balls $B[\mathbf{a}; R]$ and $B[\mathbf{b}; R]$ of radius R in \mathbb{R}^n . Then, for each \mathbf{x} in C ,

$$\begin{aligned}
 \left\| \mathbf{x} - \frac{1}{2}(\mathbf{a} + \mathbf{b}) \right\|^2 &= \|\mathbf{x}\|^2 - \mathbf{x} \cdot \mathbf{a} - \mathbf{x} \cdot \mathbf{b} + \frac{1}{4}(\|\mathbf{a}\|^2 + 2\mathbf{a} \cdot \mathbf{b} + \|\mathbf{b}\|^2) \\
 &= \frac{1}{2}(\|\mathbf{x}\|^2 - 2\mathbf{x} \cdot \mathbf{a} + \|\mathbf{a}\|^2) + \frac{1}{2}(\|\mathbf{x}\|^2 - 2\mathbf{x} \cdot \mathbf{b} + \|\mathbf{b}\|^2) \\
 &\quad - \frac{1}{4}(\|\mathbf{a}\|^2 - 2\mathbf{a} \cdot \mathbf{b} + \|\mathbf{b}\|^2) \\
 &= \frac{1}{2}\|\mathbf{x} - \mathbf{a}\|^2 + \frac{1}{2}\|\mathbf{x} - \mathbf{b}\|^2 - \frac{1}{4}\|\mathbf{a} - \mathbf{b}\|^2 \\
 &\leq R^2 - \frac{1}{4}\|\mathbf{a} - \mathbf{b}\|^2.
 \end{aligned}$$

Hence, $C \subseteq B\left[\frac{1}{2}(\mathbf{a} + \mathbf{b}); \sqrt{R^2 - \frac{1}{4}\|\mathbf{a} - \mathbf{b}\|^2}\right]$. Since C cannot lie in a closed ball of radius less than R , we must have $\mathbf{a} = \mathbf{b}$. Thus there is precisely one closed ball of radius R in \mathbb{R}^n which contains C .

Inballs, Incenters, Circumball, Circumcenter

- Let C be a convex body in \mathbb{R}^n with inradius r and circumradius R .
- Then any closed ball of radius r lying in C is called an **inball** of C and its center an **incenter** of C .
- The unique closed ball of radius R which contains C is called the **circumball** of C and its center the **circumcenter** of C .
- A (non-square) rectangle in \mathbb{R}^2 is an example of a convex body that does not have a unique incentre.

Comparisons With Elementary Geometry

- Our definitions of the terms circumradius, circumcircle and circumcenter as applied to obtuse-angled triangles do not coincide with those used in elementary geometry.
- For example, consider an isosceles triangle with sides $2, 2, 2\sqrt{3}$.
 - In the parlance of elementary geometry, its circumradius is 2 and its circumcenter lies exterior to the triangle.
 - For us here its circumradius is $\sqrt{3}$ and its circumcenter is the midpoint of its longest side.

Extremal Problems and Uniform Boundedness

- The preceding theorem asserts the existence of solutions to two extremal problems in geometry.
 - The first to find a ball of minimal radius containing a given convex body;
 - The second to find a ball of maximal radius lying in the body.
- The key step in proving the theorem was the extraction of a convergent subsequence from a sequence of closed balls.
- It is a generalization of this idea that turns out to be useful in finding solutions to many extremal problems.
- What is needed is a criterion for a sequence of sets to contain a convergent subsequence.
- A sequence of sets in \mathbb{R}^n is said to be **uniformly bounded** if there exists some ball in \mathbb{R}^n that contains every member of the sequence.

Uniform Boundedness and Cauchy Sequences

Lemma

Let A_1, \dots, A_j, \dots be a uniformly bounded sequence of nonempty compact sets in \mathbb{R}^n . Let $\varepsilon > 0$. Then there exists a subsequence $A_{i_1}, \dots, A_{i_j}, \dots$ of A_1, \dots, A_j, \dots such that $\rho(A_{i_j}, A_{i_k}) \leq \varepsilon$, for all $j, k = 1, 2, \dots$

- Since there is a ball in \mathbb{R}^n which contains every member of the given sequence, there is a finite set E in \mathbb{R}^n such that $A_j \subseteq (E)_{\frac{1}{2}\varepsilon}$, for $j = 1, 2, \dots$. For each $j = 1, 2, \dots$ denote by E_j the non-empty subset $E \cap (A_j)_{\frac{1}{2}\varepsilon}$ of E . It is easily verified that $\rho(E_j, A_j) \leq \frac{1}{2}\varepsilon$. Because E is finite, there can only be a finite number of possible different sets E_j for $j = 1, 2, \dots$. Hence the sequence E_1, \dots, E_j, \dots must contain some constant subsequence, $E_{i_1}, \dots, E_{i_j}, \dots$ say. For $j, k = 1, 2, \dots$, we have

$$\begin{aligned} \rho(A_{i_j}, A_{i_k}) &\leq \rho(A_{i_j}, E_{i_j}) + \rho(E_{i_j}, A_{i_k}) \\ &= \rho(A_{i_j}, E_{i_j}) + \rho(E_{i_k}, A_{i_k}) \leq \frac{1}{2}\varepsilon + \frac{1}{2}\varepsilon = \varepsilon. \end{aligned}$$

Uniform Boundedness and Convergence

Theorem

Every uniformly bounded sequence A_1, \dots, A_j, \dots of nonempty compact sets in \mathbb{R}^n contains a subsequence which converges to some nonempty compact set A in \mathbb{R}^n .

- It follows, by repeated applications of the lemma with $\varepsilon = 1, \frac{1}{2}, \dots, \frac{1}{j}, \dots$ that the sequence A_1, \dots, A_j, \dots contains a sequence of subsequences

$$A_{11}, A_{12}, \dots, A_{1j}, \dots$$

$$A_{21}, A_{22}, \dots, A_{2j}, \dots$$

...

$$A_{r1}, A_{r2}, \dots, A_{rj}, \dots$$

...

where each subsequence after the first in the list is a subsequence of the preceding one, and $\rho(A_{rj}, A_{rk}) \leq \frac{1}{r}$, for $j, k = 1, 2, \dots$

The diagonal sequence $A_{11}, A_{22}, \dots, A_{jj}, \dots$ is a subsequence of $A_1, A_2, \dots, A_j, \dots$ with the property that $\rho(A_{jj}, A_{kk}) \leq \frac{1}{j}$ whenever $j \leq k$.

Uniform Boundedness and Convergence (Cont'd)

- Write $B_j = A_{j,j}$ for $j = 1, 2, \dots$. We complete the proof by showing that the subsequence B_1, \dots, B_j, \dots of A_1, \dots, A_j, \dots converges to the nonempty compact set B defined by $B = \bigcap ((B_k)_{\frac{1}{k}} : k = 1, 2, \dots)$.

Let j be a positive integer and let $\mathbf{b}_j \in B_j$. For $i = 1, 2, \dots$, choose $\mathbf{b}_{j+i} \in B_{j+i}$ such that $\|\mathbf{b}_{j+i} - \mathbf{b}_j\| \leq \frac{1}{j}$; This is possible because $\rho(B_j, B_{j+i}) \leq \frac{1}{j}$. The sequence $\mathbf{b}_{j+1}, \mathbf{b}_{j+2}, \dots$ lies in the compact set $(B_1)_1$. So it contains a subsequence converging to some point \mathbf{b} of \mathbb{R}^n . Since $B_{j+i} \subseteq (B_k)_{\frac{1}{k}}$ whenever $j+i \geq k$, all but a finite number of terms of the sequence $\mathbf{b}_{j+1}, \mathbf{b}_{j+2}, \dots$ lie in the compact set $(B_k)_{\frac{1}{k}}$ for $k = 1, 2, \dots$. Hence $\mathbf{b} \in (B_k)_{\frac{1}{k}}$ and $\mathbf{b} \in B$. But \mathbf{b}_j is an arbitrary point of B_j and clearly $\|\mathbf{b} - \mathbf{b}_j\| \leq \frac{1}{j}$. So $B_j \subseteq (B)_{\frac{1}{j}}$. Trivially $B \subseteq (B_j)_{\frac{1}{j}}$. Thus, $\rho(B_j, B) \leq \frac{1}{j}$ and $B_j \rightarrow B$ as $j \rightarrow \infty$.

Blaschke Selection Principle

Theorem (Blaschke Selection Principle)

Every uniformly bounded sequence of non-empty compact convex sets in \mathbb{R}^n contains a subsequence which converges to some non-empty compact convex set in \mathbb{R}^n .

- The principle is a consequence of the theorem and the fact that a convergent sequence of convex sets must converge to a convex limit.

Continuity of Functions on Families of Compact Sets

- A typical extremal problem of elementary geometry is to maximize or minimize a real-valued function f defined on some family \mathcal{F} of nonempty compact sets in \mathbb{R}^n .
- It is important to have a concept of continuity for such functions $f : \mathcal{F} \rightarrow \mathbb{R}$.
- The function f is said to be **continuous** on \mathcal{F} if $f(A_j) \rightarrow f(A)$ as $j \rightarrow \infty$, whenever $A_j \rightarrow A$ as $j \rightarrow \infty$, where all the sets under consideration belong to \mathcal{F} .

Example

- We show that the diameter function D , which associates with each nonempty compact set A in \mathbb{R}^n its diameter $D(A)$, is continuous on the family \mathcal{F} of all nonempty compact sets in \mathbb{R}^n .
- It is easily verified that $D((A)_\lambda) = D(A) + 2\lambda$, where $A \in \mathcal{F}$ and $\lambda \geq 0$.
- Suppose now that $A, B \in \mathcal{F}$ and that $\rho(A, B) = \lambda$.
- Then

$$D(A) \leq D((B)_\lambda) = D(B) + 2\lambda \quad \text{and} \quad D(B) \leq D((A)_\lambda) = D(A) + 2\lambda.$$

- Hence $|D(A) - D(B)| \leq 2\lambda = 2\rho(A, B)$.
- The continuity of D is now clear.

An Isodiametric Problem

- We now indicate how the Blaschke selection principle can be used to show that in the family \mathcal{F} of all compact convex sets in \mathbb{R}^2 with diameter 1, there exist sets having maximal area.
- For each set A in \mathcal{F} , denote by $f(A)$ the area of A .
- This area function will be defined formally later and it will be shown to be continuous on the family of all non-empty compact convex sets in \mathbb{R}^2 .
- Let α be the supremum of the set of areas of members of \mathcal{F} .
- For each positive integer j , there is a member A_j of \mathcal{F} such that $f(A_j) > \alpha - \frac{1}{j}$.
- We may suppose, by translating the A_j 's if necessary, that they are uniformly bounded.

An Isodiametric Problem (Cont'd)

- The Blaschke selection principle guarantees the existence of a subsequence $A_{i_1}, \dots, A_{i_j}, \dots$ of A_1, \dots, A_j, \dots which converges to some non-empty compact convex set A_0 in \mathbb{R}^2 .
- The continuity of the diameter function shows that A_0 has diameter 1.
- So A_0 lies in \mathcal{F} .
- For each j ,

$$\alpha \geq f(A_{i_j}) > \alpha - \frac{1}{i_j} \geq \alpha - \frac{1}{j}.$$

- Letting $j \rightarrow \infty$ in these inequalities, we deduce, using the continuity of f , that $f(A_0) = \alpha$.
- Thus A_0 is a member of \mathcal{F} having maximal possible area.

Subsection 8

Duality

The Dual Operator

- For each nonzero vector \mathbf{u} in \mathbb{R}^n , the set $H^- = \{\mathbf{x} \in \mathbb{R}^n : \mathbf{u} \cdot \mathbf{x} \leq 1\}$ is a closed halfspace in \mathbb{R}^n containing the origin as an interior point.
- Conversely, for each closed halfspace H^- in \mathbb{R}^n containing the origin as an interior point, there is a unique non-zero vector \mathbf{u} in \mathbb{R}^n such that $H^- = \{\mathbf{x} \in \mathbb{R}^n : \mathbf{u} \cdot \mathbf{x} \leq 1\}$.
- Thus there is a bijection between the set of nonzero vectors in \mathbb{R}^n and the set of closed halfspaces in \mathbb{R}^n containing the origin as an interior point.
- We define, for each point \mathbf{u} of \mathbb{R}^n , a set \mathbf{u}^* in \mathbb{R}^n by the equation

$$\mathbf{u}^* = \{\mathbf{x} \in \mathbb{R}^n : \mathbf{u} \cdot \mathbf{x} \leq 1\}.$$

- We note that $\mathbf{0}^* = \mathbb{R}^n$.

Polar Duals of Sets

- Define the **polar dual** A^* of a set A in \mathbb{R}^n to be the intersection of all the sets \mathbf{a}^* , for $\mathbf{a} \in A$, i.e.,

$$A^* = \bigcap (\mathbf{a}^* : \mathbf{a} \in A) = \{\mathbf{x} \in \mathbb{R}^n : \mathbf{a} \cdot \mathbf{x} \leq 1, \text{ for all } \mathbf{a} \in A\}.$$

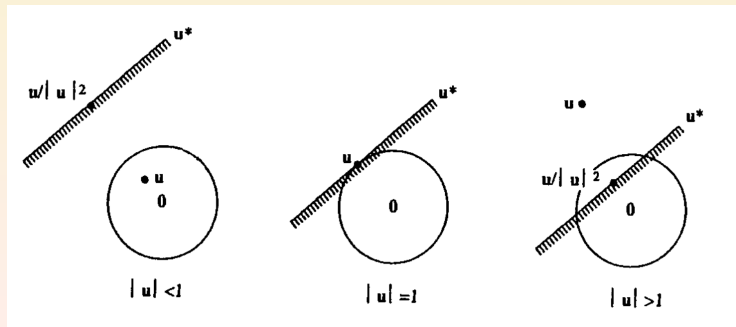
- For each set A in \mathbb{R}^n , its polar dual A^* is defined as an intersection of closed convex sets containing the origin.
- So A^* is itself a closed convex set containing the origin.
- The polar duals \emptyset^* and $\{\mathbf{0}\}^*$ are both \mathbb{R}^n .
- The polar dual of \mathbb{R}^n is $\{\mathbf{0}\}$.

Positions of Vector and its Dual

- It is instructive to examine the sets \mathbf{u}^* for nonzero vectors \mathbf{u} in \mathbb{R}^n .
- By definition, \mathbf{u}^* is the closed halfspace which is bounded by the hyperplane $H = \{\mathbf{x} \in \mathbb{R}^n : \mathbf{u} \cdot \mathbf{x} = 1\}$ and contains the origin.
- This hyperplane H has \mathbf{u} as one of its normal vectors and passes through the point $\frac{\mathbf{u}}{\|\mathbf{u}\|^2}$.
- The distance of $\frac{\mathbf{u}}{\|\mathbf{u}\|^2}$ from the origin is $\frac{1}{\|\mathbf{u}\|}$, which:
 - exceeds 1 if $\|\mathbf{u}\|$ is less than 1;
 - equals 1 if $\|\mathbf{u}\|$ equals 1;
 - is less than 1 if $\|\mathbf{u}\|$ exceeds 1.

Vector and Its Dual Illustrated

- The relative positions of \mathbf{u} and \mathbf{u}^* for the cases $\|\mathbf{u}\| < 1$, $\|\mathbf{u}\| = 1$ and $\|\mathbf{u}\| > 1$ are illustrated in the figure:



Properties of the Polar Dual

Theorem

Let A, B be sets in \mathbb{R}^n , U the closed unit ball centered on the origin of \mathbb{R}^n and λ a nonzero scalar. Then:

- (i) $A \subseteq B$ implies that $B^* \subseteq A^*$;
- (ii) $(A \cup B)^* = A^* \cap B^*$;
- (iii) $(\lambda A)^* = \frac{1}{\lambda} A^*$;
- (iv) $U^* = U$.

- (i) Suppose that $A \subseteq B$. Then

$$B^* = \{\mathbf{x} \in \mathbb{R}^n : \mathbf{b} \cdot \mathbf{x} \leq 1, \mathbf{b} \in B\} \subseteq \{\mathbf{x} \in \mathbb{R}^n : \mathbf{a} \cdot \mathbf{x} \leq 1, \mathbf{a} \in A\} = A^*.$$

- (ii) We have $\mathbf{x} \in (A \cup B)^*$ if and only if both $\mathbf{a} \cdot \mathbf{x} \leq 1$, for $\mathbf{a} \in A$, and $\mathbf{b} \cdot \mathbf{x} \leq 1$, for $\mathbf{b} \in B$, if and only if $\mathbf{x} \in A^* \cap B^*$. So $(A \cup B)^* = A^* \cap B^*$.

Properties of the Polar Dual (Cont'd)

(iii) We have

$$\begin{aligned}
 \mathbf{x} \in (\lambda A)^* & \text{ iff } \lambda \mathbf{a} \cdot \mathbf{x} \leq 1, \text{ for } \mathbf{a} \in A, \\
 & \text{ iff } \mathbf{a} \cdot (\lambda \mathbf{x}) \leq 1, \text{ for } \mathbf{a} \in A, \\
 & \text{ iff } \lambda \mathbf{x} \in A^* \\
 & \text{ iff } \mathbf{x} \in \frac{1}{\lambda} A^*.
 \end{aligned}$$

Thus $(\lambda A)^* = \frac{1}{\lambda} A^*$.

(iv) Suppose, first, that $\mathbf{x} \in U$. Then, for all $\mathbf{u} \in U$, $\mathbf{u} \cdot \mathbf{x} \leq \|\mathbf{u}\| \|\mathbf{x}\| \leq 1$. Hence $\mathbf{x} \in U^*$ and $U \subseteq U^*$.

Conversely, let $\mathbf{x} \in U^*$ with $\mathbf{x} \neq \mathbf{0}$. Then $\frac{\mathbf{x}}{\|\mathbf{x}\|} \in U$. So $\frac{\mathbf{x}}{\|\mathbf{x}\|} \cdot \mathbf{x} = \|\mathbf{x}\| \leq 1$. This shows that $\mathbf{x} \in U$. Thus $U^* \subseteq U$.

Example

- We find the polar dual A^* of the n -cube A defined by the equation

$$A = \{(a_1, \dots, a_n) : |a_1| \leq 1, \dots, |a_n| \leq 1\}.$$

- Suppose $(x_1, \dots, x_n) \in A^*$. Define $(a_1, \dots, a_n) \in A$ by $a_i = 1$, if $x_i \geq 0$, and $a_i = -1$, if $x_i < 0$. Then

$$|x_1| + \dots + |x_n| = a_1 x_1 + \dots + a_n x_n = (a_1, \dots, a_n) \cdot (x_1, \dots, x_n) \leq 1.$$

- Conversely, suppose (x_1, \dots, x_n) satisfies $|x_1| + \dots + |x_n| \leq 1$. Then, for any point $(a_1, \dots, a_n) \in A$,

$$\begin{aligned} (a_1, \dots, a_n) \cdot (x_1, \dots, x_n) &= a_1 x_1 + \dots + a_n x_n \\ &\leq |a_1| |x_1| + \dots + |a_n| |x_n| \\ &\leq |x_1| + \dots + |x_n| \leq 1. \end{aligned}$$

Thus, $(x_1, \dots, x_n) \in A^*$.

- So A^* is the set, known as a **regular n -crosspolytope**, defined by the equation $A^* = \{(x_1, \dots, x_n) : |x_1| + \dots + |x_n| \leq 1\}$.

Example (Cont'd)

- We now find A^{**} the polar dual of the polar dual A^* of the n -cube A .
- For each point (u_1, \dots, u_n) of A^{**} , define a point $(x_1, 0, \dots, 0)$ of A^* by the conditions that $x_1 = 1$ if $u_1 \geq 0$, and $x_1 = -1$ if $u_1 < 0$. Then

$$(x_1, 0, \dots, 0) \cdot (u_1, u_2, \dots, u_n) = |u_1| < 1.$$

Similarly, $|u_2| \leq 1, \dots, |u_n| \leq 1$. Hence, $(u_1, \dots, u_n) \in A$ and $A^{**} \subseteq A$.

- The inclusion $A \subseteq A^{**}$, which holds for every set A in \mathbb{R}^n , follows immediately from the definition of the polar dual.
- Hence $A^{**} = A$.

The Double-Polar Dual of a Set

- This last example suggests that we examine the **double-polar dual** A^{**} of an arbitrary set A in \mathbb{R}^n and see how it is related to A .
- The polar dual of any set in \mathbb{R}^n is always a closed convex set containing the origin.
- So a necessary condition for the equality of the sets A and A^{**} is that A is a closed convex set containing the origin.
- We aim to show that this condition is also sufficient, and establish the exact relationship between A^{**} and A .

A Set and Its Double-Polar Dual

Theorem

Let A be a set in \mathbb{R}^n . Then $A^{**} = \text{cl}(\text{conv}(A \cup \{\mathbf{0}\}))$. In particular, if A is closed, convex and contains the origin, then $A^{**} = A$.

- For all $\mathbf{a} \in A$, $\mathbf{x} \in A^*$, we have $\mathbf{a} \cdot \mathbf{x} \leq 1$. Hence $\mathbf{a} \in A^{**}$. So $A \subseteq A^{**}$. But A^{**} is a closed convex set containing $\mathbf{0}$ and A . Therefore, $\text{cl}(\text{conv}(A \cup \{\mathbf{0}\})) \subseteq A^{**}$.

For the reverse inclusion, suppose \mathbf{z} is a point of \mathbb{R}^n not lying in $\text{cl}(\text{conv}(A \cup \{\mathbf{0}\}))$. By a previous corollary, there exists a hyperplane strictly separating $\{\mathbf{z}\}$ and $\text{cl}(\text{conv}(A \cup \{\mathbf{0}\}))$. Thus, since such a hyperplane cannot pass through the origin, there exists \mathbf{u} in \mathbb{R}^n such that $\mathbf{u} \cdot \mathbf{z} > 1$ and $\mathbf{u} \cdot \mathbf{a} < 1$, for all \mathbf{a} in A . This shows that $\mathbf{u} \in A^*$ and $\mathbf{z} \notin A^{**}$. Hence, $A^{**} \subseteq \text{cl}(\text{conv}(A \cup \{\mathbf{0}\}))$.

Boundedness and the Origin

Theorem

Let A be a closed convex set in \mathbb{R}^n containing the origin. Then A is bounded if and only if the origin is an interior point of A^* , and A^* is bounded if and only if the origin is an interior point of A .

- We use two previous theorems.

Suppose first that A is bounded. Then, for some $r > 0$, $A \subseteq rU$. Hence, $\frac{1}{r}U \subseteq A^*$. So the origin is an interior point of A^* .

By applying the last result to the set A^* , we deduce that, if A^* is bounded, then the origin is an interior point of $A^{**} = A$.

Suppose next that the origin is an interior point of A . Then, for some $s > 0$, $sU \subseteq A$. Hence, $A^* \subseteq \frac{1}{s}U$. So A^* is bounded.

By applying the last result to A^* , we deduce that, if the origin is an interior point of A^* , then $A^{**} = A$ is bounded.

Polar Duals of Compacts With Origin an Interior Point

Corollary

Let \mathcal{F} be the family of all compact convex sets in \mathbb{R}^n which contain the origin as an interior point. Then the mapping $\theta: \mathcal{F} \rightarrow \mathcal{F}$ defined for $A \in \mathcal{F}$ by the equation $\theta(A) = A^*$ is a bijection.

- Let $A, B \in \mathcal{F}$. The theorem shows that $\theta(A) \in \mathcal{F}$.

θ is injective, for $A^* = B^*$ implies $A^{**} = B^{**}$, i.e., $A = B$.

It is surjective, since $\theta(A^*) = A$.

The Polar Face Mapping

- Suppose that A is a compact convex set in \mathbb{R}^n that contains the origin as an interior point.
- Let B be an exposed face of A .
- Then, for each point \mathbf{b} in B , the set $\{\mathbf{x} \in A^* : \mathbf{b} \cdot \mathbf{x} = 1\}$ is an exposed face (possibly empty) of A^* .
- Thus the set $\varphi(B)$, defined by the equation

$$\varphi(B) = \{\mathbf{x} \in A^* : \mathbf{b} \cdot \mathbf{x} = 1, \text{ for } \mathbf{b} \in B\},$$

being an intersection of exposed faces of A^* , is itself an exposed face of A^* .

- In this way we have constructed a mapping φ from the set of exposed faces of A to the set of exposed faces of A^* .
- We call φ the **polar face mapping** of A .

Properties of The Polar Face Mapping

Theorem

Let A be a compact convex set in \mathbb{R}^n which contains the origin as an interior point. Then the polar face mapping φ of A is an inclusion-reversing bijection.

- That φ is inclusion reversing follows immediately from its definition. Let ψ be the polar face mapping of A^* . We show that, for each exposed face B of A , $\psi(\varphi(B)) = B$. This is clear when B is either \emptyset or A . We assume that B is a proper exposed face of A . Thus there is \mathbf{u} in \mathbb{R}^n such that

$$B = \{\mathbf{a} \in A : \mathbf{u} \cdot \mathbf{a} = 1\} \quad \text{and} \quad \mathbf{u} \cdot \mathbf{a} \leq 1, \text{ for } \mathbf{a} \in A.$$

This shows that $\mathbf{u} \in A^*$ and $\mathbf{u} \in \varphi(B)$.

Properties of The Polar Face Mapping (Cont'd)

- Let $\mathbf{v} \in \psi(\varphi(B))$. Then $\mathbf{v} \in A^{**} = A$ and $\mathbf{u} \cdot \mathbf{v} = 1$. Hence $\mathbf{v} \in B$ and $\psi(\varphi(B)) \subseteq B$.

Conversely, let $\mathbf{b} \in B$. Then $\mathbf{b} \in A = (A^*)^*$ and $\mathbf{b} \cdot \mathbf{w} = 1$, for all $\mathbf{w} \in \varphi(B)$. Hence, $\mathbf{b} \in \psi(\varphi(B))$ and $B \subseteq \psi(\varphi(B))$.

Thus, $\psi(\varphi(B)) = B$.

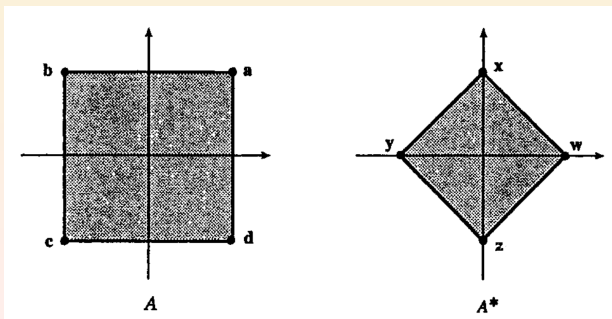
We have just shown that the composite mapping $\psi \circ \varphi$ is the identity mapping on the set of exposed faces of A .

By interchanging the roles of A and A^* in the discussion above, we can deduce that $\varphi \circ \psi$ is the identity mapping on the set of exposed faces of A^* .

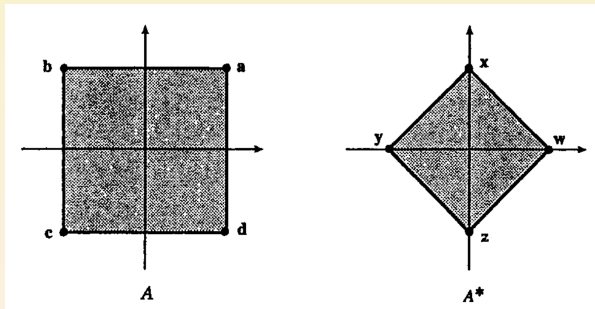
It now follows easily that φ is a bijection.

Example

- Let A be the square $\text{conv}\{\mathbf{a}, \mathbf{b}, \mathbf{c}, \mathbf{d}\}$ in \mathbb{R}^2 , where $\mathbf{a} = (1, 1)$, $\mathbf{b} = (-1, 1)$, $\mathbf{c} = (-1, -1)$, $\mathbf{d} = (1, -1)$.
- We saw that the polar dual A^* of A is the square $\text{conv}\{\mathbf{w}, \mathbf{x}, \mathbf{y}, \mathbf{z}\}$, where $\mathbf{w} = (1, 0)$, $\mathbf{x} = (0, 1)$, $\mathbf{y} = (-1, 0)$, $\mathbf{z} = (0, -1)$.



Example (Cont'd)



- The polar face mapping φ of A is indicated below, where the faces of A and A^* are represented by their extreme points.

	\emptyset	a	b	c	d	a, b	b, c	c, d	d, a	A
φ	\downarrow	\downarrow	\downarrow	\downarrow	\downarrow	\downarrow	\downarrow	\downarrow	\downarrow	\downarrow
	A^*	w, x	x, y	y, z	z, w	x	y	z	w	\emptyset