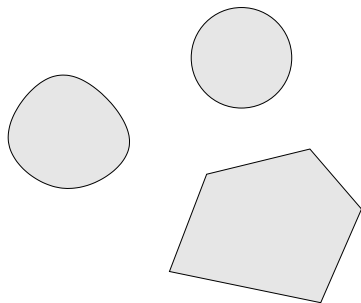


Lecture 6 - Convex Sets

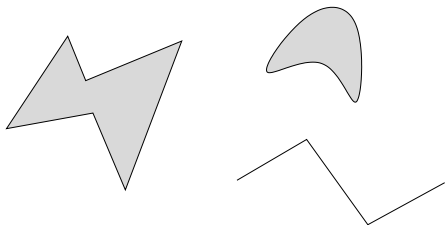
Definition A set $C \subseteq \mathbb{R}^n$ is called **convex** if for any $\mathbf{x}, \mathbf{y} \in C$ and $\lambda \in [0, 1]$, the point $\lambda\mathbf{x} + (1 - \lambda)\mathbf{y}$ belongs to C .

- ▶ The above definition is equivalent to saying that for any $\mathbf{x}, \mathbf{y} \in C$, the line segment $[\mathbf{x}, \mathbf{y}]$ is also in C .

convex sets



nonconvex sets



Examples of Convex Sets

- ▶ **Lines:** A line in \mathbb{R}^n is a set of the form

$$L = \{\mathbf{z} + t\mathbf{d} : t \in \mathbb{R}\},$$

where $\mathbf{z}, \mathbf{d} \in \mathbb{R}^n$ and $\mathbf{d} \neq \mathbf{0}$.

- ▶ $[\mathbf{x}, \mathbf{y}], (\mathbf{x}, \mathbf{y})$ for $\mathbf{x}, \mathbf{y} \in \mathbb{R}^n (\mathbf{x} \neq \mathbf{y})$.
- ▶ \emptyset, \mathbb{R}^n .
- ▶ A **hyperplane** is a set of the form

$$H = \{\mathbf{x} \in \mathbb{R}^n : \mathbf{a}^T \mathbf{x} = b\} \quad (\mathbf{a} \in \mathbb{R}^n \setminus \{\mathbf{0}\}, b \in \mathbb{R})$$

The associated **half-space** is the set

$$H^- = \{\mathbf{x} \in \mathbb{R}^n : \mathbf{a}^T \mathbf{x} \leq b\}$$

Both hyperplanes and half-spaces are convex sets.

Convexity of Balls

Lemma. Let $\mathbf{c} \in \mathbb{R}^n$ and $r > 0$. Then the open ball

$$B(\mathbf{c}, r) = \{\mathbf{x} \in \mathbb{R}^n : \|\mathbf{x} - \mathbf{c}\| < r\}$$

and the closed ball

$$B[\mathbf{c}, r] = \{\mathbf{x} \in \mathbb{R}^n : \|\mathbf{x} - \mathbf{c}\| \leq r\}$$

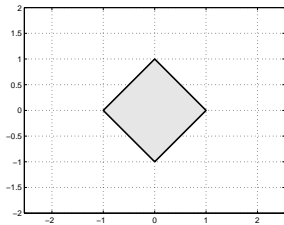
are convex.

Note that the norm is an arbitrary norm defined over \mathbb{R}^n .

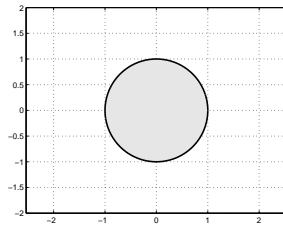
Proof. In class

l_1 , l_2 and l_∞ balls

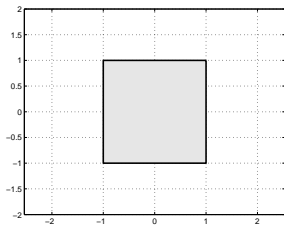
l_1



l_2



l_∞



Convexity of Ellipsoids

An **ellipsoid** is a set of the form

$$E = \{\mathbf{x} \in \mathbb{R}^n : \mathbf{x}^T \mathbf{Q} \mathbf{x} + 2\mathbf{b}^T \mathbf{x} + c \leq 0\},$$

where $\mathbf{Q} \in \mathbb{R}^{n \times n}$ is positive semidefinite, $\mathbf{b} \in \mathbb{R}^n$ and $c \in \mathbb{R}$.

Lemma: E is convex.

Proof.

- ▶ Write E as $E = \{\mathbf{x} \in \mathbb{R}^n : f(\mathbf{x}) \leq 0\}$ where $f(\mathbf{x}) \equiv \mathbf{x}^T \mathbf{Q} \mathbf{x} + 2\mathbf{b}^T \mathbf{x} + c$.

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- ▶ Take $\mathbf{x}, \mathbf{y} \in E$ and $\lambda \in [0, 1]$. Then $f(\mathbf{x}) \leq 0, f(\mathbf{y}) \leq 0$.

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$$\mathbf{z}^T \mathbf{Q} \mathbf{z} = \lambda^2 \mathbf{x}^T \mathbf{Q} \mathbf{x} + (1 - \lambda)^2 \mathbf{y}^T \mathbf{Q} \mathbf{y} + 2\lambda(1 - \lambda) \mathbf{x}^T \mathbf{Q} \mathbf{y}.$$

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- ▶

$$\begin{aligned} f(\mathbf{z}) &= \mathbf{z}^T \mathbf{Q} \mathbf{z} + 2\mathbf{b}^T \mathbf{z} + c \\ &\leq \lambda \mathbf{x}^T \mathbf{Q} \mathbf{x} + (1 - \lambda) \mathbf{y}^T \mathbf{Q} \mathbf{y} + 2\lambda \mathbf{b}^T \mathbf{x} + 2(1 - \lambda) \mathbf{b}^T \mathbf{y} + \lambda c + (1 - \lambda) c \\ &= \lambda f(\mathbf{x}) + (1 - \lambda) f(\mathbf{y}) \leq 0, \end{aligned}$$

Algebraic Operations Preserving Convexity

Lemma. Let $C_i \subseteq \mathbb{R}^n$ be a convex set for any $i \in I$ where I is an index set (possibly infinite). Then the set $\bigcap_{i \in I} C_i$ is convex.

Proof. In class

Example: Consider the set

$$P = \{\mathbf{x} \in \mathbb{R}^n : \mathbf{Ax} \leq \mathbf{b}\}$$

where $\mathbf{A} \in \mathbb{R}^{m \times n}$ and $\mathbf{b} \in \mathbb{R}^m$. P is called a **convex polyhedron** and it is indeed convex. Why?

Algebraic Operations Preserving Convexity

preservation under addition, cartesian product, forward and inverse linear mappings

Theorem.

1. Let $C_1, C_2, \dots, C_k \subseteq \mathbb{R}^n$ be convex sets and let $\mu_1, \mu_2, \dots, \mu_k \in \mathbb{R}$. Then the set $\mu_1 C_1 + \mu_2 C_2 + \dots + \mu_k C_k$ is convex.
2. Let $C_i \subseteq \mathbb{R}^{k_i}, i = 1, \dots, m$ be convex sets. Then the cartesian product

$$C_1 \times C_2 \times \dots \times C_m = \{(\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_m) : \mathbf{x}_i \in C_i, i = 1, 2, \dots, m\}$$

is convex.

3. Let $M \subseteq \mathbb{R}^n$ be a convex set and let $\mathbf{A} \in \mathbb{R}^{m \times n}$. Then the set

$$\mathbf{A}(M) = \{\mathbf{Ax} : \mathbf{x} \in M\}$$

is convex.

4. Let $D \subseteq \mathbb{R}^m$ be convex and let $\mathbf{A} \in \mathbb{R}^{m \times n}$. Then the set

$$\mathbf{A}^{-1}(D) = \{\mathbf{x} \in \mathbb{R}^n : \mathbf{Ax} \in D\}$$

is convex.

Convex Combinations

Given m points $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_m \in \mathbb{R}^n$, a **convex combination** of these m points is a vector of the form $\lambda_1 \mathbf{x}_1 + \lambda_2 \mathbf{x}_2 + \dots + \lambda_m \mathbf{x}_m$, where $\lambda_1, \lambda_2, \dots, \lambda_m$ are nonnegative numbers satisfying $\lambda_1 + \lambda_2 + \dots + \lambda_m = 1$.

- ▶ A convex set is defined by the property that any convex combination of two points from the set is also in the set.
- ▶ We will now show that a convex combination of *any* number of points from a convex set is in the set.

Convex Combinations

Theorem. Let $C \subseteq \mathbb{R}^n$ be a convex set and let $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_m \in C$. Then for any $\boldsymbol{\lambda} \in \Delta_m$, the relation $\sum_{i=1}^m \lambda_i \mathbf{x}_i \in C$ holds.

Proof by induction on m .

- ▶ For $m = 1$ the result is obvious.

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- ▶ Suppose that $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_{m+1} \in C$ and that $\boldsymbol{\lambda} \in \Delta_{m+1}$. We will show that $\mathbf{z} \equiv \sum_{i=1}^{m+1} \lambda_i \mathbf{x}_i \in C$.
- ▶ If $\lambda_{m+1} = 1$, then $\mathbf{z} = \mathbf{x}_{m+1} \in C$ and the result obviously follows.

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- ▶ If $\lambda_{m+1} < 1$ then

$$\mathbf{z} = \sum_{i=1}^m \lambda_i \mathbf{x}_i + \lambda_{m+1} \mathbf{x}_{m+1} = (1 - \lambda_{m+1}) \underbrace{\sum_{i=1}^m \frac{\lambda_i}{1 - \lambda_{m+1}} \mathbf{x}_i}_{\mathbf{v}} + \lambda_{m+1} \mathbf{x}_{m+1}.$$

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- ▶ $\mathbf{v} \in C$ and hence $\mathbf{z} = (1 - \lambda_{m+1})\mathbf{v} + \lambda_{m+1}\mathbf{x}_{m+1} \in C$.

The Convex Hull

Definition. Let $S \subseteq \mathbb{R}^n$. The **convex hull** of S , denoted by $\text{conv}(S)$, is the set comprising all the convex combinations of vectors from S :

$$\text{conv}(S) \equiv \left\{ \sum_{i=1}^k \lambda_i \mathbf{x}_i : \mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_k \in S, \boldsymbol{\lambda} \in \Delta_k \right\}.$$

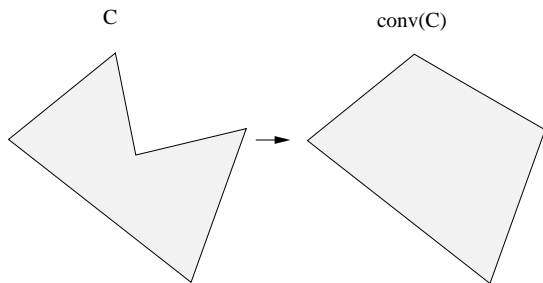


Figure: A nonconvex set and its convex hull

The Convex Hull

The convex hull $\text{conv}(S)$ is “smallest” convex set containing S .

Lemma. Let $S \subseteq \mathbb{R}^n$. If $S \subseteq T$ for some convex set T , then $\text{conv}(S) \subseteq T$.

Proof.

- ▶ Suppose that indeed $S \subseteq T$ for some convex set T .

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- ▶ There exist $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_k \in S \subseteq T$ (where k is a positive integer), and $\lambda \in \Delta_k$ such that $\mathbf{z} = \sum_{i=1}^k \lambda_i \mathbf{x}_i$.

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- ▶ Since $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_k \in T$, it follows that $\mathbf{z} \in T$, showing the desired result.

Carathéodory theorem

Theorem. Let $S \subseteq \mathbb{R}^n$ and let $\mathbf{x} \in \text{conv}(S)$. Then there exist $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_{n+1} \in S$ such that $\mathbf{x} \in \text{conv}(\{\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_{n+1}\})$, that is, there exist $\lambda \in \Delta_{n+1}$ such that

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$$\mathbf{x} = \sum_{i=1}^k \lambda_i \mathbf{x}_i.$$

- ▶ We can assume that $\lambda_i > 0$ for all $i = 1, 2, \dots, k$.
- ▶ If $k \leq n + 1$, the result is proven.

Carathéodory theorem

Theorem. Let $S \subseteq \mathbb{R}^n$ and let $\mathbf{x} \in \text{conv}(S)$. Then there exist $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_{n+1} \in S$ such that $\mathbf{x} \in \text{conv}(\{\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_{n+1}\})$, that is, there exist $\lambda \in \Delta_{n+1}$ such that

$$\mathbf{x} = \sum_{i=1}^{n+1} \lambda_i \mathbf{x}_i.$$

Proof.

- ▶ Let $\mathbf{x} \in \text{conv}(S)$. Then $\exists \mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_k \in S$ and $\lambda \in \Delta_k$ s.t.

$$\mathbf{x} = \sum_{i=1}^k \lambda_i \mathbf{x}_i.$$

- ▶ We can assume that $\lambda_i > 0$ for all $i = 1, 2, \dots, k$.
- ▶ If $k \leq n + 1$, the result is proven.
- ▶ Otherwise, if $k \geq n + 2$, then the vectors $\mathbf{x}_2 - \mathbf{x}_1, \mathbf{x}_3 - \mathbf{x}_1, \dots, \mathbf{x}_k - \mathbf{x}_1$, being more than n vectors in \mathbb{R}^n , are necessarily linearly dependent $\Rightarrow \exists \mu_2, \mu_3, \dots, \mu_k$ not all zeros s.t.

$$\sum_{i=2}^k \mu_i (\mathbf{x}_i - \mathbf{x}_1) = \mathbf{0}.$$

Proof of Carathéodory Theorem Contd.

- ▶ Defining $\mu_1 = -\sum_{i=2}^k \mu_i$, we obtain that

$$\sum_{i=1}^k \mu_i \mathbf{x}_i = \mathbf{0},$$

Proof of Carathéodory Theorem Contd.

- ▶ Defining $\mu_1 = -\sum_{i=2}^k \mu_i$, we obtain that

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- ▶ Not all of the coefficients $\mu_1, \mu_2, \dots, \mu_k$ are zeros and $\sum_{i=1}^k \mu_i = 0$.

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- ▶ Not all of the coefficients $\mu_1, \mu_2, \dots, \mu_k$ are zeros and $\sum_{i=1}^k \mu_i = 0$.
- ▶ There exists an index i for which $\mu_i < 0$. Let $\alpha \in \mathbb{R}_+$. Then

$$\mathbf{x} = \sum_{i=1}^k \lambda_i \mathbf{x}_i = \sum_{i=1}^k \lambda_i \mathbf{x}_i + \alpha \sum_{i=1}^k \mu_i \mathbf{x}_i = \sum_{i=1}^k (\lambda_i + \alpha \mu_i) \mathbf{x}_i. \quad (1)$$

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- ▶ We have $\sum_{i=1}^k (\lambda_i + \alpha \mu_i) = 1$, so (1) is a convex combination representation iff

$$\lambda_i + \alpha \mu_i \geq 0 \text{ for all } i = 1, \dots, k. \quad (2)$$

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- ▶ Since $\lambda_i > 0$ for all i , it follows that (2) is satisfied for all $\alpha \in [0, \varepsilon]$ where $\varepsilon = \min_{i:\mu_i < 0} \left\{ -\frac{\lambda_i}{\mu_i} \right\}$.

Proof of Carathéodory Theorem Contd.

- ▶ If we substitute $\alpha = \varepsilon$, then (2) still holds, but $\lambda_j + \varepsilon\mu_j = 0$ for

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- ▶ This means that we found a representation of \mathbf{x} as a convex combination of $k - 1$ (or less) vectors.

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- ▶ This means that we found a representation of \mathbf{x} as a convex combination of $k - 1$ (or less) vectors.
- ▶ This process can be carried on until a representation of \mathbf{x} as a convex combination of no more than $n + 1$ vectors is derived.

Example

For $n = 2$, consider the four vectors

$$\mathbf{x}_1 = \begin{pmatrix} 1 \\ 1 \end{pmatrix}, \mathbf{x}_2 = \begin{pmatrix} 1 \\ 2 \end{pmatrix}, \mathbf{x}_3 = \begin{pmatrix} 2 \\ 1 \end{pmatrix}, \mathbf{x}_4 = \begin{pmatrix} 2 \\ 2 \end{pmatrix},$$

and let $\mathbf{x} \in \text{conv}(\{\mathbf{x}_1, \mathbf{x}_2, \mathbf{x}_3, \mathbf{x}_4\})$ be given by

$$\mathbf{x} = \frac{1}{8}\mathbf{x}_1 + \frac{1}{4}\mathbf{x}_2 + \frac{1}{2}\mathbf{x}_3 + \frac{1}{8}\mathbf{x}_4 = \begin{pmatrix} \frac{13}{8} \\ \frac{11}{8} \end{pmatrix}.$$

Find a representation of \mathbf{x} as a convex combination of no more than 3 vectors.

In class

Convex Cones

- ▶ A set S is called a **cone** if it satisfies the following property: for any $\mathbf{x} \in S$ and $\lambda \geq 0$, the inclusion $\lambda\mathbf{x} \in S$ is satisfied.
- ▶ The following lemma shows that there is a very simple and elegant characterization of convex cones.

Lemma. A set S is a convex cone if and only if the following properties hold:

A. $\mathbf{x}, \mathbf{y} \in S \Rightarrow \mathbf{x} + \mathbf{y} \in S.$

B. $\mathbf{x} \in S, \lambda \geq 0 \Rightarrow \lambda\mathbf{x} \in S.$

Simple exercise

Examples of Convex Cones

- ▶ The convex polytope

$$C = \{\mathbf{x} \in \mathbb{R}^n : \mathbf{Ax} \leq \mathbf{0}\},$$

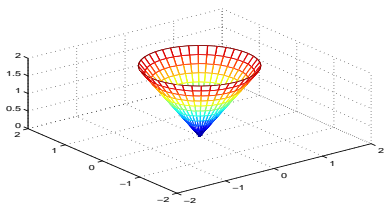
where $\mathbf{A} \in \mathbb{R}^{m \times n}$.

- ▶ **Lorentz Cone** The Lorentz cone, or *ice cream cone* is given by

$$L^n = \left\{ \begin{pmatrix} \mathbf{x} \\ t \end{pmatrix} \in \mathbb{R}^{n+1} : \|\mathbf{x}\| \leq t, \mathbf{x} \in \mathbb{R}^n, t \in \mathbb{R} \right\}.$$

- ▶ **nonnegative polynomials**. set consisting of all possible coefficients of polynomials of degree $n - 1$ which are nonnegative over \mathbb{R} :

$$K^n = \{\mathbf{x} \in \mathbb{R}^n : x_1 t^{n-1} + x_2 t^{n-2} + \dots + x_{n-1} t + x_n \geq 0 \forall t \in \mathbb{R}\}$$



The Conic Hull

Definition. Given m points $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_m \in \mathbb{R}^n$, a **conic combination** of these m points is a vector of the form $\lambda_1 \mathbf{x}_1 + \lambda_2 \mathbf{x}_2 + \dots + \lambda_m \mathbf{x}_m$, where $\lambda \in \mathbb{R}_+^m$.

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The definition of the *conic hull* is now quite natural.

Definition. Let $S \subseteq \mathbb{R}^n$. Then the **conic hull** of S , denoted by **cone**(S) is the set comprising all the conic combinations of vectors from S :

$$\text{cone}(S) \equiv \left\{ \sum_{i=1}^k \lambda_i \mathbf{x}_i : \mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_k \in S, \lambda \in \mathbb{R}_+^k \right\}.$$

Similarly to the convex hull, the conic hull of a set S is the smallest cone containing S .

Lemma. Let $S \subseteq \mathbb{R}^n$. If $S \subseteq T$ for some convex cone T , then $\text{cone}(S) \subseteq T$.

Representation Theorem for Conic Hulls

a similar result to Carathéodory theorem

Conic Representation Theorem. Let $S \subseteq \mathbb{R}^n$ and let $\mathbf{x} \in \text{cone}(S)$. Then there exist k linearly independent vector $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_k \in S$ such that $\mathbf{x} \in \text{cone}(\{\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_k\})$, that is, there exist $\boldsymbol{\lambda} \in \mathbb{R}_+^k$ such that

$$\mathbf{x} = \sum_{i=1}^k \lambda_i \mathbf{x}_i.$$

In particular, $k \leq n$.

Proof very similar to the proof of Carathéodory theorem. See page 107 of the book for the proof.

Basic Feasible Solutions

- ▶ Consider the convex polyhedron.

$$P = \{\mathbf{x} \in \mathbb{R}^n : \mathbf{Ax} = \mathbf{b}, \mathbf{x} \geq \mathbf{0}\}, \quad (\mathbf{A} \in \mathbb{R}^{m \times n}, \mathbf{b} \in \mathbb{R}^m)$$

- ▶ the rows of \mathbf{A} are assumed to be linearly independent.
- ▶ The above is a standard formulation of the constraints of a linear programming problem.

Definition. $\bar{\mathbf{x}}$ is a **basic feasible solution** (abbreviated bfs) of P if the columns of \mathbf{A} corresponding to the indices of the positive values of $\bar{\mathbf{x}}$ are linearly independent.

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Example. Consider the linear system:

$$\begin{aligned}x_1 + x_2 + x_3 &= 6 \\x_2 + x_4 &= 3 \\x_1, x_2, x_3, x_4 &\geq 0.\end{aligned}$$

Find all the basic feasible solutions. **In class**

Existence of bfs's

Theorem. Let $P = \{\mathbf{x} \in \mathbb{R}^n : \mathbf{Ax} = \mathbf{b}, \mathbf{x} \geq \mathbf{0}\}$, where $\mathbf{A} \in \mathbb{R}^{m \times n}$ and $\mathbf{b} \in \mathbb{R}^m$. If $P \neq \emptyset$, then it contains at least one bfs.

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Proof.

- ▶ $P \neq \emptyset \Rightarrow \mathbf{b} \in \text{cone}(\{\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_n\})$ where \mathbf{a}_i denotes the i -th column of \mathbf{A} .

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- ▶ By the conic representation theorem, there exist indices $i_1 < i_2 < \dots < i_k$ and k numbers $y_{i_1}, y_{i_2}, \dots, y_{i_k} \geq 0$ such that $\mathbf{b} = \sum_{j=1}^k y_{i_j} \mathbf{a}_{i_j}$ and $\mathbf{a}_{i_1}, \mathbf{a}_{i_2}, \dots, \mathbf{a}_{i_k}$ are linearly independent.

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- ▶ Denote $\bar{\mathbf{x}} = \sum_{j=1}^k y_{i_j} \mathbf{e}_{i_j}$. Then obviously $\bar{\mathbf{x}} \geq \mathbf{0}$ and in addition

$$\mathbf{A}\bar{\mathbf{x}} = \sum_{j=1}^k y_{i_j} \mathbf{A}\mathbf{e}_{i_j} = \sum_{j=1}^k y_{i_j} \mathbf{a}_{i_j} = \mathbf{b}.$$

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- ▶ Therefore, $\bar{\mathbf{x}}$ is contained in P and the columns of \mathbf{A} corresponding to the indices of the positive components of $\bar{\mathbf{x}}$ are linearly independent, meaning that P contains a bfs.

Topological Properties of Convex Sets

Theorem. Let $C \subseteq \mathbb{R}^n$ be a convex set. Then $\text{cl}(C)$ is a convex set.

Proof.

- ▶ Let $\mathbf{x}, \mathbf{y} \in \text{cl}(C)$ and let $\lambda \in [0, 1]$.

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- ▶ (**) $\lambda \mathbf{x}_k + (1 - \lambda) \mathbf{y}_k \rightarrow \lambda \mathbf{x} + (1 - \lambda) \mathbf{y}$.
- ▶ (*) + (**) $\Rightarrow \lambda \mathbf{x} + (1 - \lambda) \mathbf{y} \in \text{cl}(C)$.

The Line Segment Principle

Theorem. Let C be a convex set and assume that $\text{int}(C) \neq \emptyset$. Suppose that $\mathbf{x} \in \text{int}(C)$ and $\mathbf{y} \in \text{cl}(C)$. Then $(1 - \lambda)\mathbf{x} + \lambda\mathbf{y} \in \text{int}(C)$ for any $\lambda \in [0, 1)$.

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- ▶ Let $\mathbf{z} = (1 - \lambda)\mathbf{x} + \lambda\mathbf{y}$. We will show that $B(\mathbf{z}, (1 - \lambda)\varepsilon) \subseteq C$.

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$$\|\mathbf{w}_1 - \mathbf{y}\| < \frac{(1 - \lambda)\varepsilon - \|\mathbf{w} - \mathbf{z}\|}{\lambda}. \quad (3)$$

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$$\begin{aligned} \|\mathbf{w}_2 - \mathbf{x}\| &= \left\| \frac{\mathbf{w} - \lambda\mathbf{w}_1}{1 - \lambda} - \mathbf{x} \right\| = \frac{1}{1 - \lambda} \|(\mathbf{w} - \mathbf{z}) + \lambda(\mathbf{y} - \mathbf{w}_1)\| \\ &\leq \frac{1}{1 - \lambda} (\|\mathbf{w} - \mathbf{z}\| + \lambda\|\mathbf{w}_1 - \mathbf{y}\|) \stackrel{(3)}{<} \varepsilon, \end{aligned}$$

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Proof.

- ▶ There exists $\varepsilon > 0$ such that $B(\mathbf{x}, \varepsilon) \subseteq C$.
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- ▶ Let $\mathbf{w} \in B(\mathbf{z}, (1 - \lambda)\varepsilon)$. Since $\mathbf{y} \in \text{cl}(C)$, $\exists \mathbf{w}_1 \in C$ s.t.

$$\|\mathbf{w}_1 - \mathbf{y}\| < \frac{(1 - \lambda)\varepsilon - \|\mathbf{w} - \mathbf{z}\|}{\lambda}. \quad (3)$$

- ▶ Set $\mathbf{w}_2 = \frac{1}{1 - \lambda}(\mathbf{w} - \lambda\mathbf{w}_1)$. Then

$$\begin{aligned} \|\mathbf{w}_2 - \mathbf{x}\| &= \left\| \frac{\mathbf{w} - \lambda\mathbf{w}_1}{1 - \lambda} - \mathbf{x} \right\| = \frac{1}{1 - \lambda} \|(\mathbf{w} - \mathbf{z}) + \lambda(\mathbf{y} - \mathbf{w}_1)\| \\ &\leq \frac{1}{1 - \lambda} (\|\mathbf{w} - \mathbf{z}\| + \lambda\|\mathbf{w}_1 - \mathbf{y}\|) \stackrel{(3)}{<} \varepsilon, \end{aligned}$$

- ▶ Hence, since $B(\mathbf{x}, \varepsilon) \subseteq C$, it follows that $\mathbf{w}_2 \in C$. Finally, since $\mathbf{w} = \lambda\mathbf{w}_1 + (1 - \lambda)\mathbf{w}_2$ with $\mathbf{w}_1, \mathbf{w}_2 \in C$, we have that $\mathbf{w} \in C$.

Convexity of the Interior

Theorem. Let $C \subseteq \mathbb{R}^n$ be a convex set. Then $\text{int}(C)$ is convex.

Proof.

- ▶ If $\text{int}(C) = \emptyset$, then the theorem is obviously true.
- ▶ Otherwise, let $\mathbf{x}_1, \mathbf{x}_2 \in \text{int}(C)$, and let $\lambda \in (0, 1)$.
- ▶ By the LSP, $\lambda\mathbf{x}_1 + (1 - \lambda)\mathbf{x}_2 \in \text{int}(C)$, establishing the convexity of $\text{int}(C)$.

Combination of Closure and Interior

Lemma. Let C be a convex set with a nonempty interior. Then

1. $\text{cl}(\text{int}(C)) = \text{cl}(C)$.
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For the proof of 2, see pages 109,110 of the book for the proof of Lemma 6.30(b).

Compactness of the Convex Hull of Convex Sets

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$$\|\mathbf{y}\| = \left\| \sum_{i=1}^{n+1} \lambda_i \mathbf{x}_i \right\| \leq \sum_{i=1}^{n+1} \lambda_i \|\mathbf{x}_i\| \leq M \sum_{i=1}^{n+1} \lambda_i = M,$$

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$$\mathbf{y}_k = \sum_{i=1}^{n+1} \lambda_i^k \mathbf{x}_i^k. \tag{4}$$

Proof Contd.

- ▶ By the compactness of S and Δ_{n+1} , it follows that $\{(\boldsymbol{\lambda}^k, \mathbf{x}_1^k, \mathbf{x}_2^k, \dots, \mathbf{x}_{n+1}^k)\}_{k \geq 1}$ has a convergent subsequence $\{(\boldsymbol{\lambda}^{k_j}, \mathbf{x}_1^{k_j}, \mathbf{x}_2^{k_j}, \dots, \mathbf{x}_{n+1}^{k_j})\}_{j \geq 1}$ whose limit will be denoted by

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Example: $S = \{(0, 0)^T\} \cup \{(x, y)^T : xy \geq 1\}$

Closedness of the Conic Hull of a Finite Set

Theorem. Let $\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_k \in \mathbb{R}^n$. Then $\text{cone}(\{\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_k\})$ is closed.

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- ▶ By the conic representation theorem, each element of $\text{cone}(\{\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_k\})$ can be represented as a conic combination of a linearly independent subset of $\{\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_k\}$.

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- ▶ Therefore, if S_1, S_2, \dots, S_N are all the subsets of $\{\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_k\}$ comprising linearly independent vectors, then

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where $\mathbf{b}_1, \mathbf{b}_2, \dots, \mathbf{b}_m$ are linearly independent.

- ▶ $\text{cone}(S_i) = \{\mathbf{B}\mathbf{y} : \mathbf{y} \in \mathbb{R}_+^m\}$, where \mathbf{B} is the matrix whose columns are $\mathbf{b}_1, \mathbf{b}_2, \dots, \mathbf{b}_m$.

Proof Contd.

- ▶ Suppose that $\mathbf{x}_k \in \text{cone}(S_i)$ for all $k \geq 1$ and that $\mathbf{x}_k \rightarrow \bar{\mathbf{x}}$.

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$$\mathbf{y}_k = (\mathbf{B}^T \mathbf{B})^{-1} \mathbf{B}^T \mathbf{x}_k.$$

- ▶ Taking the limit as $k \rightarrow \infty$ in the last equation, we obtain that $\mathbf{y}_k \rightarrow \bar{\mathbf{y}}$ where $\bar{\mathbf{y}} = (\mathbf{B}^T \mathbf{B})^{-1} \mathbf{B}^T \bar{\mathbf{x}}$.

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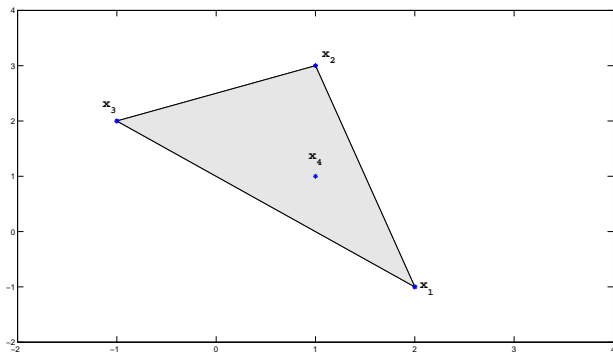
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- ▶ $\bar{\mathbf{y}} \in \mathbb{R}_+^m$.
- ▶ Thus, taking the limit in (5), we conclude that $\bar{\mathbf{x}} = \mathbf{B}\bar{\mathbf{y}}$ with $\bar{\mathbf{y}} \in \mathbb{R}_+^m$, and hence $\bar{\mathbf{x}} \in \text{cone}(S_i)$.

Extreme Points

Definition. Let $S \subseteq \mathbb{R}^n$ be a convex set. A point $\mathbf{x} \in S$ is called an **extreme point** of S if there do not exist $\mathbf{x}_1, \mathbf{x}_2 \in S (\mathbf{x}_1 \neq \mathbf{x}_2)$ and $\lambda \in (0, 1)$, such that $\mathbf{x} = \lambda \mathbf{x}_1 + (1 - \lambda) \mathbf{x}_2$.

- ▶ The set of extreme point is denoted by $\text{ext}(S)$.
- ▶ For example, the set of extreme points of a convex polytope consists of all its vertices.



Equivalence Between bfs's and Extreme Points

Theorem. Let $P = \{\mathbf{x} \in \mathbb{R}^n : \mathbf{Ax} = \mathbf{b}, \mathbf{x} \geq \mathbf{0}\}$, where $\mathbf{A} \in \mathbb{R}^{m \times n}$ has linearly independent rows and $\mathbf{b} \in \mathbb{R}^m$. The $\bar{\mathbf{x}}$ is a basic feasible solution of P if and only if it is an extreme point of P .

Theorem 6.34 in the book.

Krein-Milman Theorem

Theorem. Let $S \subseteq \mathbb{R}^n$ be a compact convex set. Then

$$S = \text{conv}(\text{ext}(S)).$$