

Introduction to Convexity

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Subsection 1

The Euclidean Space \mathbb{R}^n

Vector Space Operations in \mathbb{R}^3

- In three-dimensional coordinate geometry a **point** or **vector** is determined by its **coordinates** x, y, z relative to some rectangular coordinate system.
- We identify the point or vector with the ordered triple (x, y, z) .
- Vectors are **added** together according to a **parallelogram law**, which is equivalent to the addition of corresponding coordinates.
- The word **scalar** is used as a synonym for real number.
- The **product** of a scalar and a vector is equivalent to the multiplication of each coordinate of the vector by the scalar.
- Thus, if (x, y, z) and (u, v, w) are vectors, and λ is a scalar, then

$$\begin{aligned}(x, y, z) + (u, v, w) &= (x + u, y + v, z + w); \\ \lambda(x, y, z) &= (\lambda x, \lambda y, \lambda z).\end{aligned}$$

- These equations can be extended in the natural way to define vector addition and scalar multiplication of real n -tuples.

Euclidean Space \mathbb{R}^n

- For each positive integer n , denote by \mathbb{R}^n the set of all n -tuples (x_1, \dots, x_n) of real numbers.
- Then \mathbb{R}^n is called the n -**dimensional Euclidean space**.
- Each element $\mathbf{x} = (x_1, \dots, x_n)$ of \mathbb{R}^n is called a **point** or **vector** of \mathbb{R}^n and the real numbers x_1, \dots, x_n are called the **coordinates** of \mathbf{x} .
- For $n = 1$, we identify the 1-tuple $\mathbf{x} = (x_1)$ with the real number x_1 itself, so that \mathbb{R}^1 becomes simply \mathbb{R} , the set of real numbers.
- For $n = 1, 2, 3$, we often write x , (x, y) , (x, y, z) instead of (x_1) , (x_1, x_2) , (x_1, x_2, x_3) .
- Geometrically, \mathbb{R}^1 can be thought of as a line, \mathbb{R}^2 as a plane, and \mathbb{R}^3 as the set of points in space.
- Lower case Roman letters such as **a, b, c, x, y, z** will denote points of \mathbb{R}^n , lower case Roman and Greek letters such as $x, y, z, \lambda, \mu, \nu$ will denote scalars, and capital Roman letters such as A, B, C will denote subsets of \mathbb{R}^n .

Addition and Scalar Multiplication

- **Addition** and **scalar multiplication** in \mathbb{R}^n are defined coordinatewise.
- Thus, if $x = (x_1, \dots, x_n)$, $y = (y_1, \dots, y_n)$, and λ is a scalar, then

$$\mathbf{x} + \mathbf{y} = (x_1 + y_1, \dots, x_n + y_n) \quad \text{and} \quad \lambda \mathbf{x} = (\lambda x_1, \dots, \lambda x_n).$$

- The vector $(0, \dots, 0)$ of \mathbb{R}^n , all of whose coordinates are 0, is denoted by $\mathbf{0}$ and is called the **zero vector** or **origin** of \mathbb{R}^n .
- The vector in \mathbb{R}^n whose only non-zero coordinate is a 1 in the i th position is denoted by \mathbf{e}_i and is called the **i th elementary vector**.
- A point of \mathbb{R}^n all of whose coordinates are integers is called a **lattice point**.
- The vector $(-1)\mathbf{x}$ is written simply as $-\mathbf{x}$.
- **Vector subtraction** is defined by the rule $\mathbf{x} - \mathbf{y} = \mathbf{x} + (-1)\mathbf{y}$.
- It is sometimes convenient to write $\frac{\mathbf{x}}{\lambda}$ for $\frac{1}{\lambda}\mathbf{x}$.

\mathbb{R}^n as a Real Vector Space

- The set \mathbb{R}^n , equipped with the above operations of vector addition and scalar multiplication, is a **real vector space**.
- This means that, if $\mathbf{x}, \mathbf{y}, \mathbf{z} \in \mathbb{R}^n$ and $\lambda, \mu \in \mathbb{R}$, then the following relations hold:
 - (i) $\mathbf{x} + \mathbf{y} = \mathbf{y} + \mathbf{x}$;
 - (ii) $\mathbf{x} + (\mathbf{y} + \mathbf{z}) = (\mathbf{x} + \mathbf{y}) + \mathbf{z}$;
 - (iii) $\mathbf{x} + \mathbf{0} = \mathbf{x}$;
 - (iv) $\mathbf{x} + (-\mathbf{x}) = \mathbf{0}$;
 - (v) $1\mathbf{x} = \mathbf{x}$;
 - (vi) $\lambda(\mu\mathbf{x}) = (\lambda\mu)\mathbf{x}$;
 - (vii) $\lambda(\mathbf{x} + \mathbf{y}) = \lambda\mathbf{x} + \lambda\mathbf{y}$;
 - (viii) $(\lambda + \mu)\mathbf{x} = \lambda\mathbf{x} + \mu\mathbf{x}$.

Extending Operations on Sets

- We extend the operations of vector addition and scalar multiplication to subsets of \mathbb{R}^n by defining:

$$A + B = \{\mathbf{a} + \mathbf{b} : \mathbf{a} \in A, \mathbf{b} \in B\} \quad \text{and} \quad \lambda A = \{\lambda \mathbf{a} : \mathbf{a} \in A\},$$

where $A, B \subseteq \mathbb{R}^n$ and $\lambda \in \mathbb{R}$.

- The set $A + B$ is called the **vector sum** of A and B .
- It follows from the above definitions that both sets $A + B$ and λA are empty when A is empty.
- We write $-A$ for the set $(-1)A$, and $A - B$ for the set $A + (-B)$.
- It is sometimes convenient to write $\frac{A}{\lambda}$ for $\frac{1}{\lambda}A$.

Symmetric Sets

- The set A in \mathbb{R}^n is said to be **0-symmetric**, or simply **symmetric**, if $-A = A$.
- Geometrically, A is symmetric if it is its own reflection in the origin.
- Examples of symmetric sets in \mathbb{R}^2 are:
 - ellipses centered at the origin;
 - parallelograms with centers at the origin;
 - lines through the origin;
 - \mathbb{R}^2 itself.

Translates

- The set $\{\mathbf{a}\} + B$, where $\mathbf{a} \in \mathbb{R}^n$, is often written as $\mathbf{a} + B$ and is called a **translate** of B or, more precisely, the **translate of B by \mathbf{a}** .
- It is an easy exercise in set theory to show that

$$A + B = \bigcup (\mathbf{a} + B : \mathbf{a} \in A),$$

i.e., $A + B$ is the union of all translates of B by vectors \mathbf{a} in A .

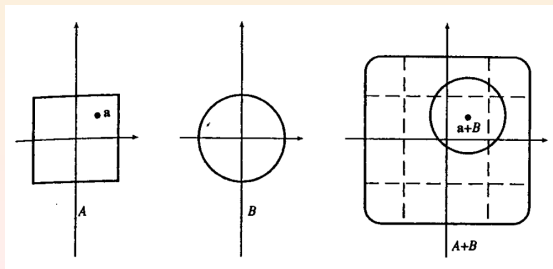
- This result can help us to visualize $A + B$ in simple cases.

Example

- Suppose that A and B are the square and the circular disc in \mathbb{R}^2 defined by the equations

$$A = \{(x, y) : |x|, |y| \leq 1\}, \quad B = \{(x, y) : x^2 + y^2 \leq 1\}.$$

- Then $\mathbf{a} + B$ is the circular disc with center \mathbf{a} and radius 1;
- $A + B$ is the union of all such discs for $\mathbf{a} \in A$.



Caution with Set Operations

- Vector addition and scalar multiplication, when applied to sets in \mathbb{R}^n , do not have all the properties one might expect, and the reader is warned to be cautious.
- For example, it is not always true that $A + A = 2A$.
To see this, let A consist of distinct points \mathbf{a} and \mathbf{b} in \mathbb{R}^n .
Then $A + A = \{2\mathbf{a}, 2\mathbf{b}, \mathbf{a} + \mathbf{b}\}$, whereas $2A = \{2\mathbf{a}, 2\mathbf{b}\}$.

Properties of Set Operations

- Properties (i)-(viii) above do, however, partially generalize to give the following easily verified results:
 - (i)* $A + B = B + A$;
 - (ii)* $A + (B + C) = (A + B) + C$;
 - (iii)* $A + \mathbf{0} = A$;
 - (iv)* $\mathbf{0} \in A + (-A)$ when $A \neq \emptyset$;
 - (v)* $1A = A$;
 - (vi)* $\lambda(\mu A) = (\lambda\mu)A$;
 - (vii)* $\lambda(A + B) = \lambda A + \lambda B$;
 - (viii)* $(\lambda + \mu)A \subseteq \lambda A + \mu A$.

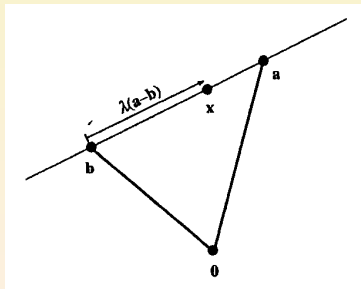
Subsection 2

Flats

Equation of a Line in \mathbb{R}^3

- For each point \mathbf{x} on the line through distinct points \mathbf{a} and \mathbf{b} of \mathbb{R}^3 , there exists a unique scalar λ such that

$$\begin{aligned}\mathbf{x} &= \mathbf{b} + \lambda(\mathbf{a} - \mathbf{b}) \\ &= \lambda\mathbf{a} + (1 - \lambda)\mathbf{b}.\end{aligned}$$



- Conversely, each point \mathbf{x} of this form lies on the line through \mathbf{a} and \mathbf{b} .
- Thus the line through \mathbf{a} and \mathbf{b} is the set $\{\lambda\mathbf{a} + (1 - \lambda)\mathbf{b} : \lambda \in \mathbb{R}\}$, which can also be written in the symmetrical form $\{\lambda\mathbf{a} + \mu\mathbf{b} : \lambda + \mu = 1\}$.
- We note that the subset

$$\{\lambda\mathbf{a} + (1 - \lambda)\mathbf{b} : 0 \leq \lambda \leq 1\} = \{\lambda\mathbf{a} + \mu\mathbf{b} : \lambda, \mu \geq 0, \lambda + \mu = 1\}$$

of the line through \mathbf{a} and \mathbf{b} is the line segment joining \mathbf{a} and \mathbf{b} .

Flats

- The **line** through distinct points \mathbf{a} and \mathbf{b} of \mathbb{R}^n is the set $\{\lambda\mathbf{a} + \mu\mathbf{b} : \lambda + \mu = 1\}$.
- Clearly this set contains both \mathbf{a} and \mathbf{b} , and its points can be placed into a bijective correspondence with the points of the real line \mathbb{R} itself.
- The set A in \mathbb{R}^n is called a **flat** if whenever it contains two points, it also contains the entire line through them.
- Expressed algebraically, A is a flat if $\lambda\mathbf{a} + \mu\mathbf{b} \in A$ whenever $\mathbf{a}, \mathbf{b} \in A$ and $\lambda + \mu = 1$.
- Equivalently, A is a flat if $\lambda A + \mu A \subseteq A$ whenever $\lambda + \mu = 1$.
- Synonyms for flat used by other authors are: **affine set**, **affine variety**, **affine manifold**, **linear variety**, and **linear manifold**.
- The empty set, singletons, lines, and \mathbb{R}^n itself are examples of flats in \mathbb{R}^n . Planes are flats in \mathbb{R}^3 .

Flats Containing the Origin

- Let A be a flat in \mathbb{R}^n which contains the origin.
- Suppose that $\mathbf{a}, \mathbf{b} \in A$ and $\lambda \in \mathbb{R}$.
- Since A is a flat and $\mathbf{a}, \mathbf{0} \in A$, $\lambda \mathbf{a} + (1 - \lambda)\mathbf{0} \in A$, i.e., $\lambda \mathbf{a} \in A$. Thus A is closed under scalar multiplication.
- Since A is a flat and $\mathbf{a}, \mathbf{b} \in A$, $\frac{1}{2}\mathbf{a} + \frac{1}{2}\mathbf{b} \in A$. But A is closed under scalar multiplication. So $2(\frac{1}{2}\mathbf{a} + \frac{1}{2}\mathbf{b}) \in A$, i.e., $\mathbf{a} + \mathbf{b} \in A$. Thus A is closed under addition.
- Hence A is a non-empty subset of \mathbb{R}^n which is closed under addition and scalar multiplication, i.e., A is a subspace of the real vector space \mathbb{R}^n .
- Trivially, a subspace of \mathbb{R}^n is a flat containing the origin.
- We have shown that flats through the origin in \mathbb{R}^n are precisely the subspaces of \mathbb{R}^n .

Relation Between Flats and Subspaces

Theorem

The non-empty flats in \mathbb{R}^n are precisely the translates of subspaces of \mathbb{R}^n .

- Suppose first that A is a non-empty flat in \mathbb{R}^n . Let $\mathbf{a} \in A$. We show that $A - \mathbf{a}$ is a flat. Let $\mathbf{x}, \mathbf{y} \in A - \mathbf{a}$ and $\lambda + \mu = 1$. Then $\lambda\mathbf{x} + \mu\mathbf{y} \in A - \mathbf{a}$. So

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

Thus, $\lambda\mathbf{x} + \mu\mathbf{y} \in A - \mathbf{a}$, and $A - \mathbf{a}$ is a flat.

Since $A - \mathbf{a}$ contains the origin, it must be a subspace of \mathbb{R}^n .

Hence the non-empty flat A is the translate of the subspace $A - \mathbf{a}$ of \mathbb{R}^n by the vector \mathbf{a} .

Relation Between Flats and Subspaces (Cont'd)

- Suppose next that A is a subspace of \mathbb{R}^n and that $\mathbf{u} \in \mathbb{R}^n$. We show that $A + \mathbf{u}$ is a flat. Let $\mathbf{x}, \mathbf{y} \in A + \mathbf{u}$ and $\lambda + \mu = 1$. Then there exist $\mathbf{a}, \mathbf{b} \in A$ such that $\mathbf{x} = \mathbf{a} + \mathbf{u}$, $\mathbf{y} = \mathbf{b} + \mathbf{u}$. So

$$\lambda \mathbf{x} + \mu \mathbf{y} = \lambda \mathbf{a} + \mu \mathbf{b} + \mathbf{u} \in A + \mathbf{u},$$

since $\lambda \mathbf{a} + \mu \mathbf{b} \in A$, as A is a subspace of \mathbb{R}^n .

This shows that $A + \mathbf{u}$ is a flat.

Uniqueness of Subspace

Corollary

Each non-empty flat in \mathbb{R}^n is the translate of precisely one subspace of \mathbb{R}^n .

- Let A be a non-empty flat in \mathbb{R}^n . Suppose that A is a translate of both the subspaces B and C of \mathbb{R}^n . Then C must be a translate of B . So there exists $\mathbf{b} \in \mathbb{R}^n$ such that $C = B + \mathbf{b}$. Since 0 lies in C , it follows that $-\mathbf{b}$, and hence \mathbf{b} , lies in B . Thus $C = B + \mathbf{b} \subseteq B$. By symmetry, $B \subseteq C$. Hence $B = C$, and A is the translate of precisely one subspace of \mathbb{R}^n .

Parallel Flats

- The observation that two (distinct) lines in \mathbb{R}^2 are parallel if and only if one is a translate of the other prompts the following definition.
- In \mathbb{R}^n a flat A is said to be **parallel** to a flat B if each is a translate of the other.
- The relation of parallelism is an equivalence relation on the family of all flats in \mathbb{R}^n .
- This notion of parallelism does not quite accord with that used in elementary geometry on two counts:
 - Firstly, a flat is considered to be parallel to itself.
 - Secondly, it only allows parallelism between flats of the same dimension. For example, we cannot speak of a line being parallel to a plane.
- The preceding corollary shows that each non-empty flat in \mathbb{R}^n is parallel to precisely one subspace of \mathbb{R}^n .

Closure Under Intersections

Theorem

The intersection of an arbitrary family of flats in \mathbb{R}^n is a flat.

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

Let $\mathbf{a}, \mathbf{b} \in \bigcap (A_i : i \in I)$ and $\lambda + \mu = 1$.

Then $\mathbf{a}, \mathbf{b} \in A_i$. As A_i is a flat, $\lambda \mathbf{a} + \mu \mathbf{b} \in A_i$, 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 a flat.

Affine Hull

- The **affine hull** $\text{aff}A$ of a set A in \mathbb{R}^n is the intersection of all flats in \mathbb{R}^n containing A .
- Such flats exist, since \mathbb{R}^n is a flat containing A .
- In view of the preceding theorem, $\text{aff}A$ is a flat which contains A .
- Moreover, if B is any flat in \mathbb{R}^n containing A , then $\text{aff}A \subseteq B$.
- Thus, we may refer to $\text{aff}A$ as the smallest flat in \mathbb{R}^n containing A .
- Clearly, A is a flat if and only if $A = \text{aff}A$.
- Moreover, $\text{aff}(\text{aff}A) = \text{aff}A$.
- Another easy result is that, if $A \subseteq B$, then $\text{aff}A \subseteq \text{aff}B$.

Affine Hull in \mathbb{R}^3

- In the space \mathbb{R}^3 :
 - The affine hull of two distinct points is the line through them;
 - The affine hull of three non-collinear points is the plane which they determine;
 - The affine hull of four non-coplanar points is the whole space \mathbb{R}^3 itself.

Generalized Flat Relation

- By definition, a set A in \mathbb{R}^n is a flat if $\lambda \mathbf{a} + \mu \mathbf{b} \in A$ whenever $\mathbf{a}, \mathbf{b} \in A$ and $\lambda + \mu = 1$.
- This defining relation of a flat implies a more general one, as we now establish in the following fundamental theorem.

Theorem

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

- Let $\mathbf{a} \in A$. Then the points $\mathbf{a}_1 - \mathbf{a}, \dots, \mathbf{a}_m - \mathbf{a}$ lie in the subspace $A - \mathbf{a}$ of \mathbb{R}^n , whence so too does the point

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

Hence $\lambda_1 \mathbf{a}_1 + \dots + \lambda_m \mathbf{a}_m \in A$.

Affine Combinations and the Affine Hull

- A point \mathbf{x} is said to be an **affine combination** of points $\mathbf{a}_1, \dots, \mathbf{a}_m$ in \mathbb{R}^n if there exist scalars $\lambda_1, \dots, \lambda_m$ 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 now be expressed as: Every affine combination of points of a flat in \mathbb{R}^n belongs to that flat.
- The affine hull of a set was defined by means of flats containing that set.
- The following theorem expresses the affine hull of a set in terms of points of the set itself.

Theorem

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

Proof

- Denote by B the set of all affine combinations of points of A . That $B \subseteq \text{aff}A$ follows from the preceding theorem and the inclusion $A \subseteq \text{aff}A$.

We next show that B is a flat. If $\mathbf{x}, \mathbf{y} \in B$, then $\mathbf{x} = \lambda_1 \mathbf{a}_1 + \cdots + \lambda_m \mathbf{a}_m$, $\mathbf{y} = \mu_1 \mathbf{b}_1 + \cdots + \mu_p \mathbf{b}_p$, for some $\mathbf{a}_1, \dots, \mathbf{a}_m, \mathbf{b}_1, \dots, \mathbf{b}_p \in A$, and scalars $\lambda_1, \dots, \lambda_m, \mu_1, \dots, \mu_p$ with $\lambda_1 + \cdots + \lambda_m = 1$, $\mu_1 + \cdots + \mu_p = 1$. Let $\lambda + \mu = 1$. 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 a flat. Since B is a flat and $B \supseteq A$, it follows that $B \supseteq \text{aff}A$. Hence $B = \text{aff}A$.

Example

Corollary

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

$$\text{aff}\{\mathbf{a}_1, \dots, \mathbf{a}_m\} = \{\lambda_1 \mathbf{a}_1 + \dots + \lambda_m \mathbf{a}_m : \lambda_1 + \dots + \lambda_m = 1\}.$$

Example: Each point $\mathbf{x} = (x_1, \dots, x_n)$ of \mathbb{R}^n can be expressed as an affine combination of the zero vector $\mathbf{0}$ and the elementary vectors $\mathbf{e}_1, \dots, \mathbf{e}_n$ as follows:

$$\mathbf{x} = (1 - x_1 - \dots - x_n)\mathbf{0} + x_1 \mathbf{e}_1 + \dots + x_n \mathbf{e}_n.$$

The corollary now shows that $\text{aff}\{\mathbf{0}, \mathbf{e}_1, \dots, \mathbf{e}_n\} = \mathbb{R}^n$.

Linear Hull

- Let A be a non-empty set in \mathbb{R}^n .
- We recall that a point 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 $\lambda_1, \dots, \lambda_m$ are scalars, is said to be a **linear combination** of points of A .
- The set of all such linear combinations is the smallest subspace of \mathbb{R}^n which contains A , and is called here the **linear hull** of A and we denote it by $\text{lin}A$.
- Since $\text{lin}A$ is a flat containing $A \cup \{\mathbf{0}\}$, it follows that $\text{aff}(A \cup \{\mathbf{0}\}) \subseteq \text{lin}A$.
- On the other hand, $\text{aff}(A \cup \{\mathbf{0}\})$ is a subspace of \mathbb{R}^n containing A , so $\text{lin}A \subseteq \text{aff}(A \cup \{\mathbf{0}\})$.
- We conclude that $\text{lin}A = \text{aff}(A \cup \{\mathbf{0}\})$.
- We define $\text{lin}\emptyset = \{\mathbf{0}\}$.
- This ensures that $\text{lin}\emptyset$ is the smallest subspace of \mathbb{R}^n which contains \emptyset , and that $\text{lin}\emptyset = \text{aff}(\emptyset \cup \{\mathbf{0}\})$.

Addition and Scalar Multiplication

- We conclude the section by examining how flats behave with respect to the operations of addition and scalar multiplication.

Theorem

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

- Let $\lambda + \mu = 1$. Since A and B are flats, $\lambda A + \mu A \subseteq A$ and $\lambda B + \mu B \subseteq B$. Thus,

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

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

Corollary

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

Scalar Distributivity

- We saw in the last section that it is not in general true that $A + A = 2A$.
- It is true, however, when A is a flat.

Theorem

Let A be a flat in \mathbb{R}^n and let $\lambda_1, \dots, \lambda_m$ be scalars with $\lambda_1 + \dots + \lambda_m \neq 0$. Then

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

- Write $\lambda = \lambda_1 + \dots + \lambda_m$. Then, using a previous theorem, we 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 $(\lambda_1 + \dots + \lambda_m)A = \lambda_1 A + \dots + \lambda_m A$.

Subsection 3

Dimension

Affine Dependence

- The set A in \mathbb{R}^n is said to be **affinely dependent** if there exists $\mathbf{a} \in A$ such that $\mathbf{a} \in \text{aff}(A \setminus \{\mathbf{a}\})$.
- Thus in \mathbb{R}^3 :
 - A set of three points is affinely dependent if and only if it is collinear;
 - A set of four points is affinely dependent if and only if it is coplanar;
 - Any set having more than four points is affinely dependent.

Affine Independence

- A set in \mathbb{R}^n which is not affinely dependent is said to be **affinely independent**.
- In \mathbb{R}^3 :
 - A set of three points is affinely independent precisely when it is the vertex set of a non-degenerate triangle;
 - A set of four points is affinely independent precisely when it is the vertex set of a non-degenerate tetrahedron.
- In \mathbb{R}^n , the empty set, every singleton, and every set consisting of two points are affinely independent.
- Since any set in \mathbb{R}^n which contains an affinely dependent set is itself affinely dependent, it follows that every subset of an affinely independent set is affinely independent.

Criterion for Affine Dependence

Theorem

Let A be a set in \mathbb{R}^n . Then A is affinely dependent if and only if there exist distinct points $\mathbf{a}_1, \dots, \mathbf{a}_m$ of A and scalars $\lambda_1, \dots, \lambda_m$, not all zero, such that

$$\lambda_1 \mathbf{a}_1 + \dots + \lambda_m \mathbf{a}_m = \mathbf{0} \quad \text{and} \quad \lambda_1 + \dots + \lambda_m = 0.$$

- Suppose that A is affinely dependent. Then there exists $\mathbf{a}_1 \in A$ such that $\mathbf{a}_1 \in \text{aff}(A \setminus \{\mathbf{a}_1\})$. By a previous theorem, there exist (distinct) points $\mathbf{a}_2, \dots, \mathbf{a}_m$ of $A \setminus \{\mathbf{a}_1\}$ and scalars μ_2, \dots, μ_m , such that $\mathbf{a}_1 = \mu_2 \mathbf{a}_2 + \dots + \mu_m \mathbf{a}_m$ and $\mu_2 + \dots + \mu_m = 1$. Write $\lambda_1 = -1$, $\lambda_2 = \mu_2$, \dots , $\lambda_m = \mu_m$. Then $\lambda_1, \dots, \lambda_m$ are not all zero and satisfy the conclusion.

Criterion for Affine Dependence (Cont'd)

- Suppose next that there exist distinct points $\mathbf{a}_1, \dots, \mathbf{a}_m$ of A , and scalars $\lambda_1, \dots, \lambda_m$, not all zero, which satisfy the hypothesis.

Suppose that $\lambda_1 \neq 0$. Then

$$\mathbf{a}_1 = -\frac{1}{\lambda_1}(\lambda_2 \mathbf{a}_2 + \dots + \lambda_m \mathbf{a}_m) \quad \text{and} \quad -\frac{1}{\lambda_1}(\lambda_2 + \dots + \lambda_m) = 1,$$

which shows that \mathbf{a}_1 is an affine combination of $\mathbf{a}_2, \dots, \mathbf{a}_m$. Hence $\mathbf{a}_1 \in \text{aff}\{\mathbf{a}_2, \dots, \mathbf{a}_m\} \subseteq \text{aff}(A \setminus \{\mathbf{a}_1\})$. So A is affinely dependent.

Corollary

A subset $\{\mathbf{a}_1, \dots, \mathbf{a}_m\}$ of \mathbb{R}^n is affinely dependent if and only if there exist scalars $\lambda_1, \dots, \lambda_m$, not all zero, such that

$$\lambda_1 \mathbf{a}_1 + \dots + \lambda_m \mathbf{a}_m = \mathbf{0} \quad \text{and} \quad \lambda_1 + \dots + \lambda_m = 0.$$

Uniqueness

Corollary

Let $\{\mathbf{a}_1, \dots, \mathbf{a}_m\}$ be an affinely independent set in \mathbb{R}^n . Then each point of $\text{aff}\{\mathbf{a}_1, \dots, \mathbf{a}_m\}$ can be expressed uniquely in the form

$$\lambda_1 \mathbf{a}_1 + \dots + \lambda_m \mathbf{a}_m, \quad \text{where} \quad \lambda_1 + \dots + \lambda_m = 1.$$

- A previous corollary shows that each point of $\text{aff}\{\mathbf{a}_1, \dots, \mathbf{a}_m\}$ can be expressed in the desired form.

To establish the uniqueness, suppose that

$$\begin{aligned} \lambda_1 \mathbf{a}_1 + \dots + \lambda_m \mathbf{a}_m &= \mu_1 \mathbf{a}_1 + \dots + \mu_m \mathbf{a}_m, \\ \lambda_1 + \dots + \lambda_m &= \mu_1 + \dots + \mu_m = 1. \end{aligned}$$

Uniqueness (Cont'd)

- Then

$$(\lambda_1 - \mu_1)\mathbf{a}_1 + \cdots + (\lambda_m - \mu_m)\mathbf{a}_m = \mathbf{0}$$

with $(\lambda_1 - \mu_1) + \cdots + (\lambda_m - \mu_m) = 0$.

Since $\{\mathbf{a}_1, \dots, \mathbf{a}_m\}_{\neq}$ is affinely independent, the preceding corollary shows that the scalars $\lambda_1 - \mu_1, \dots, \lambda_m - \mu_m$ must be zero.

Thus $\lambda_1 = \mu_1, \dots, \lambda_m = \mu_m$, and the uniqueness is established.

Cardinality of Affinely Independent Sets

- We mentioned that any set of more than four points in \mathbb{R}^3 is affinely dependent.

Corollary

An affinely independent set in \mathbb{R}^n cannot contain more than $n+1$ points.

- It suffices to show that every set of the form $\{\mathbf{a}_1, \dots, \mathbf{a}_m\} \neq \emptyset$ in \mathbb{R}^n , where $m > n+1$, is affinely dependent. Let $\{\mathbf{a}_1, \dots, \mathbf{a}_m\} \neq \emptyset$ be a set in \mathbb{R}^n , where $m > n+1$. Then the system of the $n+1$ linear simultaneous equations

$$\lambda_1 \mathbf{a}_1 + \dots + \lambda_m \mathbf{a}_m = \mathbf{0}, \quad \lambda_1 + \dots + \lambda_m = 0,$$

in the m unknowns $\lambda_1, \dots, \lambda_m$ is homogeneous. Since $m > n+1$, it has a non-trivial solution. Hence, $\{\mathbf{a}_1, \dots, \mathbf{a}_m\} \neq \emptyset$, is affinely dependent by a previous corollary.

Affine Hull and Affine Independence

Corollary

Let A be an affinely independent subset of \mathbb{R}^n . Suppose that \mathbf{a} is a point of \mathbb{R}^n not lying in $\text{aff}A$. Then the set $A \cup \{\mathbf{a}\}$ is affinely independent.

- We argue by contradiction. Suppose that $A \cup \{\mathbf{a}\}$ is affinely dependent. Then there exist distinct points $\mathbf{a}_1, \dots, \mathbf{a}_m$ of A and scalars $\lambda, \lambda_1, \dots, \lambda_m$, not all zero, such that $\lambda \mathbf{a} + \lambda_1 \mathbf{a}_1 + \dots + \lambda_m \mathbf{a}_m = \mathbf{0}$ and $\lambda + \lambda_1 + \dots + \lambda_m = 0$. The scalar λ cannot be zero, for then A is affinely dependent. Thus the equation can be used to express \mathbf{a} as an affine combination of $\mathbf{a}_1, \dots, \mathbf{a}_m$. So $\mathbf{a} \in \text{aff}\{\mathbf{a}_1, \dots, \mathbf{a}_m\}$. This, however, contradicts the hypothesis that $\mathbf{a} \notin \text{aff}A$. Hence $A \cup \{\mathbf{a}\}$ is affinely independent.

Example

- In \mathbb{R}^n the set $\{\mathbf{0}, \mathbf{e}_1, \dots, \mathbf{e}_n\}$ is affinely independent.
- To see this, suppose that the scalars $\lambda, \lambda_1, \dots, \lambda_n$ satisfy

$$\lambda \mathbf{0} + \lambda_1 \mathbf{e}_1 + \dots + \lambda_n \mathbf{e}_n = \mathbf{0} \quad \text{and} \quad \lambda + \lambda_1 + \dots + \lambda_n = 0.$$

The first of these equations shows that $\lambda_1, \dots, \lambda_n$ are all zero. Hence λ must also be zero from the second equation.

The corollary now shows that the set $\{\mathbf{0}, \mathbf{e}_1, \dots, \mathbf{e}_n\}$ is affinely independent.

Independent Generators

- In \mathbb{R}^3 as a simple case-by-case consideration shows, each r -dimensional flat ($r = 0, 1, 2, 3$) is the affine hull of some affinely independent set of $r + 1$ points.
- For example, a plane is the affine hull of any three of its points which are not collinear.
- Previous examples show that \mathbb{R}^3 is the affine hull of the affinely independent set $\{\mathbf{0}, \mathbf{e}_1, \mathbf{e}_2, \mathbf{e}_3\}$.
- This suggests that we might assign a *dimension* r to a flat in \mathbb{R}^n if it is the affine hull of some affinely independent set of $r + 1$ points.
- Before we can formalize this idea, however, two results need to be established:
 - (i) Every flat in \mathbb{R}^n is the affine hull of some finite affinely independent set;
 - (ii) If two affinely independent sets in \mathbb{R}^n have the same affine hull, then they have the same number of elements.

Dimension Theorem

Theorem

Every flat in \mathbb{R}^n is the affine hull of some finite affinely independent subset of \mathbb{R}^n . Moreover, the number of elements in such a subset is determined uniquely by the flat itself.

- Consider the non-trivial case of a flat A in \mathbb{R}^n which is neither empty nor a singleton. Let m be the largest positive integer such that A contains an affinely independent subset of $m+1$ elements. Such an m exists by a previous corollary, and $m \geq 1$, since A contains at least two points. Let $\{\mathbf{a}_0, \mathbf{a}_1, \dots, \mathbf{a}_m\}$ be an affinely independent subset of A . Since A is a flat, $\text{aff}\{\mathbf{a}_0, \mathbf{a}_1, \dots, \mathbf{a}_m\} \subseteq A$. Now $A \subseteq \text{aff}\{\mathbf{a}_0, \mathbf{a}_1, \dots, \mathbf{a}_m\}$, for otherwise there would exist some point \mathbf{a} of A not lying in $\text{aff}\{\mathbf{a}_0, \mathbf{a}_1, \dots, \mathbf{a}_m\}$ and, by a previous corollary, $\{\mathbf{a}, \mathbf{a}_0, \mathbf{a}_1, \dots, \mathbf{a}_m\}$ would be an affinely independent subset of A having $m+2$ elements, so contradicting the definition of m . Hence $A = \text{aff}\{\mathbf{a}_0, \mathbf{a}_1, \dots, \mathbf{a}_m\}$.

Dimension Theorem (Cont'd)

- We now complete the proof by showing that m is the dimension of the unique subspace B of \mathbb{R}^n that is parallel to A .

This we do by proving that the subset $\{\mathbf{a}_1 - \mathbf{a}_0, \dots, \mathbf{a}_m - \mathbf{a}_0\}$ of B is a basis for B . Let $\mathbf{b} \in B$. Then $\mathbf{b} = \mathbf{x} - \mathbf{a}_0$ for some $\mathbf{x} \in A$. Thus, there exist scalars $\lambda_0, \lambda_1, \dots, \lambda_m$ such that

$$\mathbf{x} = \lambda_0 \mathbf{a}_0 + \lambda_1 \mathbf{a}_1 + \dots + \lambda_m \mathbf{a}_m$$

and $\lambda_0 + \lambda_1 + \dots + \lambda_m = 1$. Hence,

$$\mathbf{b} = \mathbf{x} - \mathbf{a}_0 = \lambda_1 (\mathbf{a}_1 - \mathbf{a}_0) + \dots + \lambda_m (\mathbf{a}_m - \mathbf{a}_0).$$

This shows that $\{\mathbf{a}_1 - \mathbf{a}_0, \dots, \mathbf{a}_m - \mathbf{a}_0\}$ spans B .

Dimension Theorem (Cont'd)

- Finally, suppose that μ_1, \dots, μ_m satisfy

$$\mu_1(\mathbf{a}_1 - \mathbf{a}_0) + \dots + \mu_m(\mathbf{a}_m - \mathbf{a}_0) = \mathbf{0}.$$

Then

$$\begin{aligned} -(\mu_1 + \dots + \mu_m)\mathbf{a}_0 + \mu_1\mathbf{a}_1 + \dots + \mu_m\mathbf{a}_m &= \mathbf{0}, \\ -(\mu_1 + \dots + \mu_m) + \mu_1 + \dots + \mu_m &= 0. \end{aligned}$$

But $\{\mathbf{a}_0, \mathbf{a}_1, \dots, \mathbf{a}_m\} \neq$ is affinely independent. So all of μ_1, \dots, μ_m are zero. Thus $\{\mathbf{a}_1 - \mathbf{a}_0, \dots, \mathbf{a}_m - \mathbf{a}_0\}$ is linearly independent. We conclude that $\{\mathbf{a}_1 - \mathbf{a}_0, \dots, \mathbf{a}_m - \mathbf{a}_0\}$ is a basis for B .

Hence, m is the dimension of B , and so is uniquely determined by A .

Dimension of Flats

- A flat in \mathbb{R}^n which is the affine hull of some affinely independent set of $r+1$ points is said to have **dimension** r and is called an **r -flat**.
- It follows from the theorem that each flat in \mathbb{R}^n has a unique dimension r attached to it, and from a previous corollary that $r \leq n$.
- The empty flat is the affine hull of the (affinely independent) empty set, and so has dimension -1 .
- Clearly every singleton (point) has dimension 0 and every line has dimension 1.
- We have already seen that \mathbb{R}^n is the affine hull of the affinely independent set $\{\mathbf{0}, \mathbf{e}_1, \dots, \mathbf{e}_n\}$, whence \mathbb{R}^n has dimension n .

Dimension of Subsets

- The concept of dimension is extended to arbitrary subsets of \mathbb{R}^n by defining the **dimension** $\dim A$ of a set A in \mathbb{R}^n to be the dimension of the flat $\text{aff} A$.
- We note that when a flat in \mathbb{R}^n is also a subspace of \mathbb{R}^n its dimension as defined above coincides with its dimension as a subspace of the real vector space \mathbb{R}^n .
- Hence we may apply the term dimension unambiguously both to flats and subspaces of \mathbb{R}^n .

Dimension Equation

Theorem

Let A and B be flats in \mathbb{R}^n which have a non-empty intersection. Then

$$\dim(A + B) + \dim(A \cap B) = \dim A + \dim B.$$

- Let $\mathbf{c} \in A \cap B$. Then $A - \mathbf{c}$ and $B - \mathbf{c}$ are subspaces of \mathbb{R}^n . So, by the dimension theorem of elementary linear algebra, $\dim((A - \mathbf{c}) + (B - \mathbf{c})) + \dim((A - \mathbf{c}) \cap (B - \mathbf{c})) = \dim(A - \mathbf{c}) + \dim(B - \mathbf{c})$, that is, $\dim(A + B - 2\mathbf{c}) + \dim((A \cap B) - \mathbf{c}) = \dim(A - \mathbf{c}) + \dim(B - \mathbf{c})$. The proof of the preceding theorem shows that the dimension of a non-empty flat in \mathbb{R}^n coincides with the dimension of the unique subspace of \mathbb{R}^n which is parallel to it. It follows from this last result that the dimension of any translate of a flat is the same as the dimension of the flat itself. Thus, the last equation above simplifies to $\dim(A + B) + \dim(A \cap B) = \dim A + \dim B$.

Affine Bases

- An **affine basis** for a flat in \mathbb{R}^n is any affinely independent set in \mathbb{R}^n whose affine hull is that flat.
- A previous theorem shows that every flat has an affine basis.
- By definition, every affine basis for an r -flat has precisely $r + 1$ elements.

Example: $\{\mathbf{0}, \mathbf{e}_1, \dots, \mathbf{e}_n\}$ is an affine basis for \mathbb{R}^n .

- The next result shows that every affinely independent subset of a set in \mathbb{R}^n can be extended to an affine basis for the affine hull of the set.

Extension to an Affine Basis

Theorem

Let B be an affinely independent subset of a set A in \mathbb{R}^n . Then there exists an affine basis for $\text{aff}A$ that lies in A and contains B .

- Consider the non-empty family \mathcal{F} of all affinely independent subsets of A which contain B . Since no affinely independent set in \mathbb{R}^n contains more than $n+1$ points, there must exist some member C of \mathcal{F} that is not properly contained in any other member of \mathcal{F} . Since C is a subset of A , we have $\text{aff}C \subseteq \text{aff}A$. We claim that $\text{aff}C = \text{aff}A$.

Suppose that $\text{aff}C \subset \text{aff}A$. Since $\text{aff}A$ is the smallest flat containing A , we cannot have $A \subseteq \text{aff}C$, whence there exists some point \mathbf{a} of A not lying in $\text{aff}C$. We can now use a previous corollary to deduce that $C \cup \{\mathbf{a}\}$ is a member of \mathcal{F} which properly contains C . This contradicts the choice of C . Thus $\text{aff}C = \text{aff}A$ and C is an affine basis of $\text{aff}A$.

Barycentric Coordinates

Corollary

Let A be a subset of \mathbb{R}^n . Then A contains an affine basis for $\text{aff}A$.

- Let $\{\mathbf{a}_0, \dots, \mathbf{a}_r\}$ be an affine basis for a non-empty r -flat A in \mathbb{R}^n .
- Then, by a previous corollary, each point \mathbf{x} of A can be expressed uniquely in the form

$$\mathbf{x} = \lambda_0 \mathbf{a}_0 + \dots + \lambda_r \mathbf{a}_r, \quad \text{where } \lambda_0 + \dots + \lambda_r = 1.$$

- The scalars $\lambda_0, \dots, \lambda_r$ are called the **barycentric coordinates** of \mathbf{x} relative to (the ordered affine basis) $\mathbf{a}_0, \dots, \mathbf{a}_r$.
- A previous example shows that the barycentric coordinates of a point $\mathbf{x} = (x_1, \dots, x_n)$ of \mathbb{R}^n relative to $\mathbf{0}, \mathbf{e}_1, \dots, \mathbf{e}_n$ are $1 - x_1 - \dots - x_n, x_1, \dots, x_n$.

Scalars of Point Relative to Affine Basis

Theorem

Let $\{\mathbf{a}_0, \dots, \mathbf{a}_r\}$ be an affine basis for a non-empty r -flat A in \mathbb{R}^n . Let $\lambda_0, \dots, \lambda_r$ be the barycentric coordinates of a point $\mathbf{x} = (x_1, \dots, x_n)$ of A relative to $\mathbf{a}_0, \dots, \mathbf{a}_r$. Then there exist scalars a_{ij} ($i = 0, \dots, r, j = 0, \dots, n$) such that, for $i = 0, \dots, r$,

$$\lambda_i = a_{i0} + a_{i1}x_1 + \dots + a_{in}x_n.$$

- Extend, if necessary, $\{\mathbf{a}_0, \dots, \mathbf{a}_r\}$ to an affine basis $\{\mathbf{a}_0, \dots, \mathbf{a}_n\}$ for \mathbb{R}^n . Each point $\mathbf{x} = (x_1, \dots, x_n)$ of \mathbb{R}^n can be written uniquely in the form

$$\mathbf{x} = \lambda_0 \mathbf{a}_0 + \dots + \lambda_n \mathbf{a}_n, \quad \text{where } \lambda_0 + \dots + \lambda_n = 1.$$

In particular, each of the points $\mathbf{0}, \mathbf{e}_1, \dots, \mathbf{e}_n$ can be so expressed.

Write $\mathbf{e}_0 = \mathbf{0}$.

Scalars of Point (Cont'd)

- Then there are scalars b_{ij} ($i = 0, \dots, n; j = 0, \dots, n$) such that, for $i = 0, \dots, n$,

$$\mathbf{e}_i = b_{0i}\mathbf{a}_0 + \dots + b_{ni}\mathbf{a}_n \quad \text{and} \quad b_{0i} + \dots + b_{ni} = 1.$$

Write $\mathbf{x} = (1 - x_1 - \dots - x_n)\mathbf{e}_0 + x_1\mathbf{e}_1 + \dots + x_n\mathbf{e}_n$.

Then $\mathbf{x} = \mu_0\mathbf{a}_0 + \dots + \mu_n\mathbf{a}_n$, where, for $i = 0, \dots, n$,

$$\mu_i = b_{i0}(1 - x_1 - \dots - x_n) + b_{i1}x_1 + \dots + b_{in}x_n.$$

A routine verification shows that $\mu_0 + \dots + \mu_n = 1$.

Since the representation of \mathbf{x} in this form is unique, we can deduce that $\lambda_i = \mu_i$, for $i = 0, \dots, n$.

We complete the proof by putting $a_{i0} = b_{i0}$ for $i = 0, \dots, n$, and $a_{ij} = b_{ij} - b_{i0}$ ($i = 0, \dots, n, j = 1, \dots, n$), and noting that $\mathbf{x} \in A$ if and only if $\lambda_{r+1} = 0, \dots, \lambda_n = 0$.

Non-Meetings 1-Flats

Theorem

Let L and M be two lines that lie in a 2-flat A of \mathbb{R}^n and which do not meet. Then L and M are parallel.

- Let \mathbf{a}, \mathbf{b} be distinct points of L , and let \mathbf{c}, \mathbf{d} be distinct points of M . Since $\{\mathbf{a}, \mathbf{b}, \mathbf{c}\}$ is affinely independent, it will form an affine basis for A . Thus $\mathbf{d} = \alpha\mathbf{a} + \beta\mathbf{b} + \gamma\mathbf{c}$ for some $\alpha, \beta, \gamma \in \mathbb{R}$ with $\alpha + \beta + \gamma = 1$. A typical point on M , the line joining \mathbf{c} and \mathbf{d} , has the form

$$(1 - \theta)\mathbf{c} + \theta\mathbf{d} = \theta\alpha\mathbf{a} + \theta\beta\mathbf{b} + (\theta(\gamma - 1) + 1)\mathbf{c},$$

for some $\theta \in \mathbb{R}$. Since the latter point does not lie on L for any θ , we must have $\gamma = 1$ and $\mathbf{d} = \alpha(\mathbf{a} - \mathbf{b}) + \mathbf{c}$. Hence $\mathbf{d} - (\mathbf{c} - \mathbf{a}) = \alpha(\mathbf{a} - \mathbf{b}) + \mathbf{c} - \mathbf{c} + \mathbf{a} = (\alpha + 1)\mathbf{a} - \alpha\mathbf{b} \in L$. Thus, $M - (\mathbf{c} - \mathbf{a}) \subseteq L$. Since $M - (\mathbf{c} - \mathbf{a})$ is a line, we must have $M - (\mathbf{c} - \mathbf{a}) = L$. Thus L and M are parallel.

Subsection 4

Hyperplanes

Linear Equations

- Consider the following system of m linear equations in n real variables x_1, \dots, x_n :

$$\begin{cases} a_{11}x_1 + \dots + a_{1n}x_n = a_{10} \\ \vdots \\ a_{m1}x_1 + \dots + a_{mn}x_n = a_{m0} \end{cases}$$

where a_{ij} are given scalars.

- By the **solution set** of this system is meant the set of all n -tuples (x_1, \dots, x_n) of \mathbb{R}^n that satisfy it.
- The solution set of the system is clearly the intersection of the solution sets of the m linear equations which comprise it.

Solution Sets and Hyperplanes

- An easy verification shows that the solution set of any one of the individual linear equations is a flat.
- So the solution set of the whole system is a flat.
- Later in the section, we shall show that every flat is the solution set of some system of linear equations.
- In general, the solution set of a single linear equation $a_1x_1 + \cdots + a_nx_n = a_0$ is an $(n-1)$ -dimensional flat in \mathbb{R}^n .
- In the study of convexity in \mathbb{R}^n , flats of dimension $n-1$ play a key role, and are given their own name, **hyperplanes**.

Hyperplanes

- To be precise, we should refer not to a *hyperplane*, but to a *hyperplane in \mathbb{R}^n* .
- When no ambiguity is likely to arise, however, we do speak simply of a hyperplane.
- A hyperplane:
 - in \mathbb{R}^1 is a point;
 - in \mathbb{R}^2 is a line;
 - in \mathbb{R}^3 is a plane.
- Thus:
 - A hyperplane in \mathbb{R}^2 has an equation of the form $ax + by + c = 0$, where not both of a and b are zero;
 - A hyperplane in \mathbb{R}^3 has an equation of the form $ax + by + cz + d = 0$, where not all of a , b and c are zero.

Characterization of Hyperplanes

Theorem

A set H in \mathbb{R}^n is a hyperplane if and only if there exist scalars c_0, c_1, \dots, c_n , where not all c_1, \dots, c_n are zero, such that

$$H = \{(x_1, \dots, x_n) : c_0 + c_1x_1 + \dots + c_nx_n = 0\}.$$

- Let $H = \{(x_1, \dots, x_n) : c_0 + c_1x_1 + \dots + c_nx_n = 0\}$, where c_0, c_1, \dots, c_n are scalars and not all c_1, \dots, c_n are zero, say $c_1 \neq 0$. Let $\mathbf{u} = (u_1, \dots, u_n)$, $\mathbf{v} = (v_1, \dots, v_n)$ lie in H and let $\lambda + \mu = 1$. Then

$$\begin{aligned} & c_0 + c_1(\lambda u_1 + \mu v_1) + \dots + c_n(\lambda u_n + \mu v_n) \\ &= \lambda(c_0 + c_1 u_1 + \dots + c_n u_n) + \mu(c_0 + c_1 v_1 + \dots + c_n v_n) \\ &= \lambda 0 + \mu 0 = 0. \end{aligned}$$

Thus $\lambda \mathbf{u} + \mu \mathbf{v} \in H$ and H is a flat.

Characterization of Hyperplanes (Cont'd)

- Define points $\mathbf{a}_1, \dots, \mathbf{a}_n$ of H by the equations $\mathbf{a}_1 = (-\frac{c_0}{c_1}, 0, 0, \dots, 0)$ and $\mathbf{a}_2 = (-\frac{c_0+c_2}{c_1}, 1, 0, \dots, 0), \dots, \mathbf{a}_n = (-\frac{c_0+c_n}{c_1}, 0, 0, \dots, 1)$. Since H is a flat, $\text{aff}\{\mathbf{a}_1, \dots, \mathbf{a}_n\} \subseteq H$.

We now establish the opposite inclusion. Let $\mathbf{x} \in H$. Then the equations

$$\mathbf{x} = (x_1, \dots, x_n) = (1 - x_2 - \dots - x_n)\mathbf{a}_1 + x_2\mathbf{a}_2 + \dots + x_n\mathbf{a}_n$$

express \mathbf{x} as an affine combination of $\mathbf{a}_1, \dots, \mathbf{a}_n$. So $\mathbf{x} \in \text{aff}\{\mathbf{a}_1, \dots, \mathbf{a}_n\}$. Hence, $H \subseteq \text{aff}\{\mathbf{a}_1, \dots, \mathbf{a}_n\}$ and, therefore, $H = \text{aff}\{\mathbf{a}_1, \dots, \mathbf{a}_n\}$.

Characterization of Hyperplanes (Cont'd)

- To show that the set $\{\mathbf{a}_1, \dots, \mathbf{a}_n\}$ is affinely independent, suppose that $\lambda_1, \dots, \lambda_n$ satisfy

$$\lambda_1 \mathbf{a}_1 + \dots + \lambda_n \mathbf{a}_n = \mathbf{0} \quad \text{and} \quad \lambda_1 + \dots + \lambda_n = 0.$$

Comparing the i th coordinates ($i = 2, \dots, n$) on both sides of the first of these equations, we find that $\lambda_2, \dots, \lambda_n$ are all zero. Thus, so too is λ_1 , from the second equation.

So, $\{\mathbf{a}_1, \dots, \mathbf{a}_n\}$ is affinely independent.

But $H = \text{aff}\{\mathbf{a}_1, \dots, \mathbf{a}_n\}$, and so H is an $(n-1)$ -dimensional flat, i.e., H is a hyperplane.

Characterization of Hyperplanes (Converse)

- Conversely, suppose that H is a hyperplane in \mathbb{R}^n . Let $\{\mathbf{b}_1, \dots, \mathbf{b}_n\}$ be an affine basis for H . Extend this to an affine basis $\{\mathbf{b}_0, \mathbf{b}_1, \dots, \mathbf{b}_n\}$ for \mathbb{R}^n . Then each $\mathbf{x} = (x_1, \dots, x_n)$ in \mathbb{R}^n can be written uniquely in the form

$$\mathbf{x} = \lambda_0 \mathbf{b}_0 + \lambda_1 \mathbf{b}_1 + \dots + \lambda_n \mathbf{b}_n, \quad \text{where } \lambda_0 + \lambda_1 + \dots + \lambda_n = 1.$$

Thus $\lambda_0, \lambda_1, \dots, \lambda_n$ are the barycentric coordinates of \mathbf{x} relative to the (ordered) affine basis $\mathbf{b}_0, \mathbf{b}_1, \dots, \mathbf{b}_n$. By a previous theorem, there exist scalars c_0, c_1, \dots, c_n such that

$$\lambda_0 = c_0 + c_1 x_1 + \dots + c_n x_n.$$

Since $\mathbf{x} \in H$ iff $\lambda_0 = 0$, $H = \{(x_1, \dots, x_n) : c_0 + c_1 x_1 + \dots + c_n x_n = 0\}$. Not all of c_1, \dots, c_n are zero, for this would imply that either H is empty (if $c_0 \neq 0$) or \mathbb{R}^n (if $c_0 = 0$), both of which contradict the assumption that H is an $(n-1)$ -dimensional flat.

Characterization of r -Flats

Corollary

In \mathbb{R}^n each r -flat ($r = -1, \dots, n$) can be expressed as the intersection of $n - r$ hyperplanes, and so is the solution set of some system of $n - r$ linear equations.

- The only (-1) -flat in \mathbb{R}^n is the empty set, which is the intersection of the $n + 1$ hyperplanes $x_1 = 0, \dots, x_n = 0, x_1 + \dots + x_n = 1$.

The only n -flat in \mathbb{R}^n is \mathbb{R}^n itself, which is the intersection of no hyperplanes.

Consider now the case of an r -flat A in \mathbb{R}^n , where $r = 0, \dots, n - 1$. Let $\{\mathbf{a}_0, \dots, \mathbf{a}_r\}$ be an affine basis for A . Extend this to an affine basis $\{\mathbf{a}_0, \dots, \mathbf{a}_n\}$ for \mathbb{R}^n . Then each \mathbf{x} in \mathbb{R}^n can be expressed uniquely in the form

$$\mathbf{x} = \lambda_0 \mathbf{a}_0 + \dots + \lambda_n \mathbf{a}_n, \quad \text{where} \quad \lambda_0 + \dots + \lambda_n = 1.$$

Characterization of r -Flats (Cont'd)

- Now A is the set in \mathbb{R}^n consisting precisely of those \mathbf{x} 's whose barycentric coordinates $\lambda_{r+1}, \dots, \lambda_n$ are all zero.

But each of the sets $\{\mathbf{x} : \lambda_i = 0\}$ is the hyperplane

$$\text{aff}\{\mathbf{a}_0, \dots, \mathbf{a}_{i-1}, \mathbf{a}_{i+1}, \dots, \mathbf{a}_n\}.$$

It now follows that A is the intersection of the $n - r$ hyperplanes with equations $\lambda_{r+1} = 0, \dots, \lambda_n = 0$.

Uniqueness of Constants

- Given a hyperplane H in \mathbb{R}^n , there exist scalars c_0, c_1, \dots, c_n , with not all c_1, \dots, c_n zero, such that

$$H = \{(x_1, \dots, x_n) : c_0 + c_1x_1 + \dots + c_nx_n = 0\}.$$

- We now consider to what extent H determines the scalars c_0, c_1, \dots, c_n .
- It certainly does not determine them uniquely, for the scalars $\theta c_0, \theta c_1, \dots, \theta c_n$, where $\theta \neq 0$, serve equally well in the equation for H .
- Suppose that d_0, d_1, \dots, d_n are also scalars such that

$$H = \{(x_1, \dots, x_n) : d_0 + d_1x_1 + \dots + d_nx_n = 0\}.$$

- Assume that $c_1 \neq 0$, and let $\mathbf{a}_1, \dots, \mathbf{a}_n$ be the points of H specified as in the first part of the proof of the preceding theorem.

Uniqueness of Constants (Cont'd)

- Substituting the coordinates of the \mathbf{a}_i into the above equation for H in terms of the d 's, we deduce that $d_i = \frac{d_1}{c_1} c_i$ for $i = 0, \dots, n$.
- Since not all of d_0, d_1, \dots, d_n can be zero, we deduce that d_1 , and hence $\frac{d_1}{c_1}$, is not zero.
- Writing $\theta = \frac{d_1}{c_1}$, we find that $d_0 = \theta c_0, d_1 = \theta c_1, \dots, d_n = \theta c_n$.
- Thus the hyperplane H determines the scalars c_0, c_1, \dots, c_n to within a common non-zero scalar multiple.

Halfspaces

- The importance of hyperplanes in \mathbb{R}^n is that they divide the whole space into two halfspaces in a natural way.

Example: A line in \mathbb{R}^2 with equation $ax + by + c = 0$ divides \mathbb{R}^2 into the two halfplanes determined by the inequalities $ax + by + c \leq 0$ and $ax + by + c \geq 0$.

- A hyperplane in \mathbb{R}^n with equation $c_0 + c_1x_1 + \cdots + c_nx_n = 0$ divides \mathbb{R}^n into the two halfspaces determined by the inequalities

$$c_0 + c_1x_1 + \cdots + c_nx_n \leq 0 \quad \text{and} \quad c_0 + c_1x_1 + \cdots + c_nx_n \geq 0.$$

- Let c_0, c_1, \dots, c_n be scalars, where not all c_1, \dots, c_n are zero. Then a set of either of the forms

$$\{(x_1, \dots, x_n) : c_0 + c_1x_1 + \cdots + c_nx_n \leq 0\} \quad \text{or} \\ \{(x_1, \dots, x_n) : c_0 + c_1x_1 + \cdots + c_nx_n \geq 0\}$$

is called a **closed halfspace** in \mathbb{R}^n .

Halfspaces Determined by a Hyperplane

- A set of either of the forms

$$\{(x_1, \dots, x_n) : c_0 + c_1x_1 + \dots + c_nx_n < 0\} \quad \text{or} \\ \{(x_1, \dots, x_n) : c_0 + c_1x_1 + \dots + c_nx_n > 0\}$$

is called a **open halfspace** in \mathbb{R}^n .

- If the scalars c_0, c_1, \dots, c_n are replaced, respectively, by $\theta c_0, \theta c_1, \dots, \theta c_n$, for some $\theta \neq 0$, then we obtain the same pair of closed halfspaces and the same pair of open halfspaces, although the order of the halfspaces is reversed when $\theta < 0$.
- Thus, if H is a hyperplane in \mathbb{R}^n with equation $c_0 + c_1x_1 + \dots + c_nx_n = 0$, then the above pair of closed halfspaces and the above pair of open halfspaces are determined by H (independent of equation).
- Hence we may refer unambiguously to the closed halfspaces and the open halfspaces **determined by H** .
- We say that the closed (open) halfspaces determined by H are **opposite** to one another.

Example

- Any line through two points lying in opposite halfspaces determined by a hyperplane in \mathbb{R}^n meets the hyperplane.
- Suppose that the hyperplane H has equation $c_0 + c_1x_1 + \cdots + c_nx_n = 0$, and that the points $\mathbf{a} = (a_1, \dots, a_n)$ and $\mathbf{b} = (b_1, \dots, b_n)$ lie in opposite halfspaces determined by H .
- Omitting the trivial case when either of \mathbf{a} or \mathbf{b} lies on H ,

$$c_0 + c_1a_1 + \cdots + c_na_n = \alpha < 0 \quad \text{and} \quad c_0 + c_1b_1 + \cdots + c_nb_n = \beta > 0.$$

- The points on the line L through \mathbf{a} and \mathbf{b} are precisely those points of the form $(\lambda a_1 + (1 - \lambda)b_1, \dots, \lambda a_n + (1 - \lambda)b_n)$, where the scalar λ assumes all real values.
- We find, by substituting these coordinates into the equation of H , that $\lambda = \frac{\beta}{\beta - \alpha}$ corresponds to the unique point of intersection of L and H .
- This value of λ satisfies $0 \leq \lambda \leq 1$. So the portion of L lying, between \mathbf{a} and \mathbf{b} , the so-called **line segment** joining \mathbf{a} and \mathbf{b} , meets H .

Characterization of Parallel Hyperplanes

Theorem

Let H and H' be hyperplanes in \mathbb{R}^n with respective equations $c_0 + c_1x_1 + \cdots + c_nx_n = 0$ and $c'_0 + c'_1x_1 + \cdots + c'_nx_n = 0$. Then H and H' are parallel if and only if there exists a scalar θ such that $c'_1 = \theta c_1, \dots, c'_n = \theta c_n$.

- Suppose first that H and H' are parallel, say $H' = H + \mathbf{a}$, where $\mathbf{a} = (a_1, \dots, a_n)$. Then $(x_1, \dots, x_n) \in H$ if and only if

$$\begin{aligned} & c'_0 + c'_1(x_1 + a_1) + \cdots + c'_n(x_n + a_n) \\ &= c'_0 + c'_1a_1 + \cdots + c'_na_n + c'_1x_1 + \cdots + c'_nx_n = 0. \end{aligned}$$

Thus, by the above remarks on the representation of hyperplanes by linear equations, there exists a θ , such that $c'_1 = \theta c_1, \dots, c'_n = \theta c_n$.

Characterization of Parallel Hyperplanes

- Suppose next that $c'_1 = \theta c_1, \dots, c'_n = \theta c_n$, where θ is a (non-zero) scalar.

Then, for $d_0 = \frac{c'_0}{\theta}$, H' is represented by the equation

$$d_0 + c_1 x_1 + \cdots + c_n x_n = 0.$$

Let $\mathbf{b} = (b_1, \dots, b_n)$ satisfy

$$c_1 b_1 + \cdots + c_n b_n = c_0 - d_0.$$

Then H' also has the equation

$$c_0 + c_1(x_1 - b_1) + \cdots + c_n(x_n - b_n) = 0.$$

Thus $\mathbf{x} = (x_1, \dots, x_n) \in H'$ if and only if $\mathbf{x} - \mathbf{b} \in H$.

Hence $H' = H + \mathbf{b}$. This shows that H and H' are parallel.

Relative Position of Hyperplanes

Corollary

Two parallel hyperplanes in \mathbb{R}^n are either identical or disjoint. Two non-parallel hyperplanes in \mathbb{R}^n must meet.

- Let H and H' be parallel hyperplanes in \mathbb{R}^n . Then they have respective equations

$$c_0 + c_1x_1 + \cdots + c_nx_n = 0 \quad \text{and} \quad c'_0 + \theta c_1x_1 + \cdots + \theta c_nx_n = 0,$$

say, where θ is a non-zero scalar. If $c'_0 = \theta c_0$, then H and H' are identical. Otherwise they are disjoint.

Relative Position of Hyperplanes (Cont'd)

- Let H and H' be non-parallel hyperplanes in \mathbb{R}^n having respective equations

$$c_0 + c_1x_1 + \cdots + c_nx_n = 0 \quad \text{and} \quad c'_0 + c'_1x_1 + \cdots + c'_nx_n = 0.$$

Then there is no scalar θ such that $c'_1 = \theta c_1, \dots, c'_n = \theta c_n$. It follows that $n \geq 2$. Suppose that $c_1 \neq 0$. Then, for some $j \in \{2, \dots, n\}$, $c'_j \neq \frac{c'_1}{c_1}c_j$, say $c'_2 \neq \frac{c'_1}{c_1}c_2$. It is easily verified that the point

$$\left(\frac{c'_0c_2 - c_0c'_2}{c_1c'_2 - c'_1c_2}, \frac{c_0c'_1 - c'_0c_1}{c_1c'_2 - c'_1c_2}, 0, \dots, 0 \right)$$

lies in $H \cap H'$. So H and H' meet.

Subsection 5

Affine Transformations

Affine Transformations

- A mapping $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is called an **affine transformation** if $T(\lambda \mathbf{x} + \mu \mathbf{y}) = \lambda T(\mathbf{x}) + \mu T(\mathbf{y})$ whenever $\mathbf{x}, \mathbf{y} \in \mathbb{R}^n$ and $\lambda + \mu = 1$.

- A simple example of an affine transformation is the mapping $T : \mathbb{R}^3 \rightarrow \mathbb{R}^3$ defined by the equation $T(x, y, z) = (x, y, 1)$.

Geometrically, T is the orthogonal projection of \mathbb{R}^3 onto the plane with equation $z = 1$.

- For each vector $\mathbf{q} \in \mathbb{R}^n$, the mapping $T : \mathbb{R}^n \rightarrow \mathbb{R}^n$ defined by the equation $T(\mathbf{x}) = \mathbf{x} + \mathbf{q}$ is an affine transformation called the **translation** of \mathbb{R}^n through \mathbf{q} .

Affine versus Linear Transformations

- Clearly every linear transformation from \mathbb{R}^n to \mathbb{R}^m is also an affine one.
- That not every affine transformation from \mathbb{R}^n to \mathbb{R}^m is linear, follows from the observation that it need not map the zero vector of \mathbb{R}^n to the zero vector of \mathbb{R}^m .
- See the two examples of affine transformations given above.
- The exact relationship between linear and affine transformations is given in the following result.

Relation Between Affine and Linear Transformations

Theorem

Let $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$ be an affine transformation. Then T is linear if and only if $T(\mathbf{0}) = \mathbf{0}$.

- In view of the remarks above, it will suffice to show that T is linear when $T(\mathbf{0}) = \mathbf{0}$.

Suppose, then, that $T(\mathbf{0}) = \mathbf{0}$. Let $\mathbf{x}, \mathbf{y} \in \mathbb{R}^n$ and $\lambda \in \mathbb{R}$. Then

$$T(\lambda \mathbf{x}) = T(\lambda \mathbf{x} + (1 - \lambda)\mathbf{0}) = \lambda T(\mathbf{x}) + (1 - \lambda)T(\mathbf{0}) = \lambda T(\mathbf{x}).$$

Using this last result, we deduce that

$$\begin{aligned} T(\mathbf{x} + \mathbf{y}) &= T\left(2\left(\frac{1}{2}\mathbf{x} + \frac{1}{2}\mathbf{y}\right)\right) = 2T\left(\frac{1}{2}\mathbf{x} + \frac{1}{2}\mathbf{y}\right) \\ &= 2\left(\frac{1}{2}T(\mathbf{x}) + \frac{1}{2}T(\mathbf{y})\right) = T(\mathbf{x}) + T(\mathbf{y}). \end{aligned}$$

Thus T is linear.

Matrix Form of an Affine Transformation

- In the following discussion, all vectors considered will be identified with column vectors in the natural way.

Theorem

The affine transformations $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$ are precisely those mappings $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$ which can be expressed in the form $T(\mathbf{x}) = \mathbf{Q}\mathbf{x} + \mathbf{q}$, for some real $m \times n$ matrix \mathbf{Q} and some real $m \times 1$ matrix \mathbf{q} .

- It is easily verified that a mapping of the type under consideration is an affine transformation.

Assume, then, that $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is an affine transformation. Let $T(\mathbf{0}) = \mathbf{q}$. Then the mapping $T' : \mathbb{R}^n \rightarrow \mathbb{R}^m$ defined by the equation $T'(\mathbf{x}) = T(\mathbf{x}) - \mathbf{q}$ is readily shown to be an affine transformation with $T'(\mathbf{0}) = \mathbf{0}$. The theorem shows that T' is linear, whence there is a real $m \times n$ matrix \mathbf{Q} such that $T'(\mathbf{x}) = \mathbf{Q}\mathbf{x}$. Thus $T(\mathbf{x}) = \mathbf{Q}\mathbf{x} + \mathbf{q}$.

Remarks

- The affine transformation $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$ determines the matrices \mathbf{Q} and \mathbf{q} uniquely:
 - The j th column of \mathbf{Q} must be $T(\mathbf{e}_j) - T(\mathbf{0})$;
 - \mathbf{q} must be $T(\mathbf{0})$.
- The above representation of an affine transformation in terms of matrices shows easily that, if $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is an affine transformation, $\mathbf{a}_1, \dots, \mathbf{a}_r \in \mathbb{R}^n$ and $\lambda_1 + \dots + \lambda_r = 1$, then

$$T(\lambda_1 \mathbf{a}_1 + \dots + \lambda_r \mathbf{a}_r) = \lambda_1 T(\mathbf{a}_1) + \dots + \lambda_r T(\mathbf{a}_r).$$

Affine Transformations and Flats

Corollary

Let $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$ be an affine transformation and let A be a set in \mathbb{R}^n . Then $T(\text{aff}A) = \text{aff}T(A)$. If A is a flat, then so too is $T(A)$.

- A point \mathbf{x} lies in $T(\text{aff}A)$ if and only if there exist $\mathbf{a}_1, \dots, \mathbf{a}_r \in A$ and $\lambda_1, \dots, \lambda_r$ with $\lambda_1 + \dots + \lambda_r = 1$ such that

$$\mathbf{x} = T(\lambda_1 \mathbf{a}_1 + \dots + \lambda_r \mathbf{a}_r) = \lambda_1 T(\mathbf{a}_1) + \dots + \lambda_r T(\mathbf{a}_r),$$

that is, if and only if $\mathbf{x} \in \text{aff}T(A)$. Thus $T(\text{aff}A) = \text{aff}T(A)$.

If A is a flat, then $\text{aff}T(A) = T(\text{aff}A) = T(A)$. This shows that $T(A)$ is a flat.

Non-Singular Affine Transformations

- Consider an affine transformation $T : \mathbb{R}^n \rightarrow \mathbb{R}^n$ of \mathbb{R}^n into itself.
- By the theorem, there exist a real $n \times n$ matrix \mathbf{Q} and a real $n \times 1$ matrix \mathbf{q} such that $T(\mathbf{x}) = \mathbf{Q}\mathbf{x} + \mathbf{q}$.
- The affine transformation T is said to be **non-singular** if the determinant $\det \mathbf{Q}$ of the matrix \mathbf{Q} is non-zero, that is if \mathbf{Q} has an inverse, i.e., is non-singular.

Invertible Affine Transformations

Theorem

Let $T : \mathbb{R}^n \rightarrow \mathbb{R}^n$ be an affine transformation. Then T has an inverse if and only if T is non-singular. When T is non-singular, its inverse $T^{-1} : \mathbb{R}^n \rightarrow \mathbb{R}^n$ is an affine transformation.

- Let \mathbf{Q} be a real $n \times n$ matrix and \mathbf{q} a real $n \times 1$ matrix such that $T(\mathbf{x}) = \mathbf{Q}\mathbf{x} + \mathbf{q}$ for all \mathbf{x} in \mathbb{R}^n . Suppose first that \mathbf{Q} is non-singular. Then $\det \mathbf{Q}$ is non-zero and \mathbf{Q} has an inverse \mathbf{Q}^{-1} . For each \mathbf{y} in \mathbb{R}^n , the equation $T(\mathbf{x}) = \mathbf{y}$ has the unique solution $\mathbf{x} = \mathbf{Q}^{-1}\mathbf{y} - \mathbf{Q}^{-1}\mathbf{q}$.

It follows that T has an inverse, which is the affine transformation $T^{-1} : \mathbb{R}^n \rightarrow \mathbb{R}^n$ defined by the equation $T^{-1}(\mathbf{y}) = \mathbf{Q}^{-1}\mathbf{y} - \mathbf{Q}^{-1}\mathbf{q}$ for \mathbf{y} in \mathbb{R}^n .

Suppose next that $\det \mathbf{Q}$ is zero. Then there exists a non-zero vector \mathbf{z} in \mathbb{R}^n such that $\mathbf{Q}(\mathbf{z}) = \mathbf{0}$. Hence $T(\mathbf{z}) = T(\mathbf{0})$ and T is not injective. Hence T has no inverse.

Affine Transformations and Affinely Independent Sets

Theorem

Let $\{\mathbf{a}_0, \dots, \mathbf{a}_r\} \neq$ and $\{\mathbf{b}_0, \dots, \mathbf{b}_r\} \neq$ be affinely independent sets in \mathbb{R}^n . Then there exists a non-singular affine transformation $T: \mathbb{R}^n \rightarrow \mathbb{R}^n$ such that $T(\mathbf{a}_i) = \mathbf{b}_i$, for $i = 0, \dots, r$.

- Extend the sets $\{\mathbf{a}_0, \dots, \mathbf{a}_r\} \neq$ and $\{\mathbf{b}_0, \dots, \mathbf{b}_r\} \neq$, respectively, to affine bases $\{\mathbf{a}_0, \dots, \mathbf{a}_n\}$ and $\{\mathbf{b}_0, \dots, \mathbf{b}_n\}$ for \mathbb{R}^n . Then each \mathbf{x} in \mathbb{R}^n can be written uniquely in the form $\mathbf{x} = \lambda_0 \mathbf{a}_0 + \dots + \lambda_n \mathbf{a}_n$, $\lambda_0 + \dots + \lambda_n = 1$.

Define a mapping $T: \mathbb{R}^n \rightarrow \mathbb{R}^n$ by the equation

$$T(\mathbf{x}) = \lambda_0 \mathbf{b}_0 + \dots + \lambda_n \mathbf{b}_n.$$

It is routine to verify that T is a bijective affine transformation. Hence T is a non-singular affine transformation such that $T(\mathbf{a}_i) = \mathbf{b}_i$, for $i = 0, \dots, r$.

Affine Transformations and Flats

Corollary

Let A and B be flats in \mathbb{R}^n of the same dimension. Then there is a non-singular affine transformation $T : \mathbb{R}^n \rightarrow \mathbb{R}^n$ such that $T(A) = B$.

- If A and B are both empty, then T can be taken as the identity mapping of \mathbb{R}^n onto itself.

Suppose, then, that A and B are non-empty and have affine bases $\{\mathbf{a}_0, \dots, \mathbf{a}_r\} \neq \emptyset$ and $\{\mathbf{b}_0, \dots, \mathbf{b}_r\} \neq \emptyset$, respectively.

Let T be as in the theorem. Then, by a previous corollary,

$$T(A) = T(\text{aff}\{\mathbf{a}_0, \dots, \mathbf{a}_r\}) = \text{aff}\{\mathbf{b}_0, \dots, \mathbf{b}_r\} = B.$$

r -Dimensional Flats in \mathbb{R}^n

- Suppose that B is an r -dimensional flat ($r \geq 1$) in \mathbb{R}^n and that $A = \text{aff}\{\mathbf{0}, \mathbf{e}_1, \dots, \mathbf{e}_r\}$.
- Then A and B are flats of the same dimension.
- By the corollary, there exists a non-singular affine transformation $T: \mathbb{R}^n \rightarrow \mathbb{R}^n$ such that $T(A) = B$.
- The flat A consists precisely of those points (x_1, \dots, x_n) for which $x_{r+1} = 0, \dots, x_n = 0$.
- Hence A can be identified with \mathbb{R}^r by associating the point (x_1, \dots, x_n) of A with the point (x_1, \dots, x_r) of \mathbb{R}^r .
- Under this identification $T(\mathbb{R}^r) = B$.
- Thus every r -dimensional flat ($r \geq 1$) can be considered to be an affine copy of \mathbb{R}^r .
- This identification is often helpful when working with r -dimensional sets in \mathbb{R}^n , for we may consider them as subsets of \mathbb{R}^r and make use of the resulting algebraic simplification.

Subsection 6

Length, Distance and Angle

The Inner Product

- The **inner product** $\mathbf{x} \cdot \mathbf{y}$ of vectors $\mathbf{x} = (x_1, \dots, x_n)$ and $\mathbf{y} = (y_1, \dots, y_n)$ in \mathbb{R}^n is the real number defined by the equation

$$\mathbf{x} \cdot \mathbf{y} = x_1 y_1 + \cdots + x_n y_n.$$

- The following properties of the inner product are immediate consequences of its definition.
- For $\mathbf{x}, \mathbf{y}, \mathbf{z} \in \mathbb{R}^n$ and $\lambda, \mu \in \mathbb{R}$:
 - (i) $\mathbf{x} \cdot \mathbf{x} \geq 0$, and $\mathbf{x} \cdot \mathbf{x} = 0$ if and only if $\mathbf{x} = \mathbf{0}$;
 - (ii) $\mathbf{x} \cdot \mathbf{y} = \mathbf{y} \cdot \mathbf{x}$;
 - (iii) $(\lambda \mathbf{x} + \mu \mathbf{y}) \cdot \mathbf{z} = \lambda(\mathbf{x} \cdot \mathbf{z}) + \mu(\mathbf{y} \cdot \mathbf{z})$.

The Norm and the Distance

- The **norm** or **length** $\|\mathbf{x}\|$ of a vector $\mathbf{x} = (x_1, \dots, x_n)$ in \mathbb{R}^n is the non-negative real number defined by the equation

$$\|\mathbf{x}\| = \sqrt{\mathbf{x} \cdot \mathbf{x}}, \quad \text{whence} \quad \|\mathbf{x}\| = \sqrt{x_1^2 + \dots + x_n^2}.$$

- The **distance** between points $\mathbf{x} = (x_1, \dots, x_n)$ and $\mathbf{y} = (y_1, \dots, y_n)$ of \mathbb{R}^n is the non-negative real number

$$\|\mathbf{x} - \mathbf{y}\| = \sqrt{(x_1 - y_1)^2 + \dots + (x_n - y_n)^2},$$

i.e., the length of the vector $\mathbf{x} - \mathbf{y}$, or $\mathbf{y} - \mathbf{x}$.

Properties of the Norm

- The norm of the zero vector is 0.
- The norm of each elementary vector \mathbf{e}_i is 1.
- In general, any vector in \mathbb{R}^n which has norm 1 is called a **unit vector**.
- The following properties of the norm are simple consequences of its definition.
- For $\mathbf{x}, \mathbf{y} \in \mathbb{R}^n$ and $\lambda, \mu \in \mathbb{R}$:
 - $\|\mathbf{x}\| \geq 0$, and $\|\mathbf{x}\| = 0$ if and only if $\mathbf{x} = \mathbf{0}$;
 - $\|\lambda\mathbf{x}\| = |\lambda|\|\mathbf{x}\|$;
 - $\|\lambda\mathbf{x} + \mu\mathbf{y}\|^2 = \lambda^2\|\mathbf{x}\|^2 + 2\lambda\mu\mathbf{x} \cdot \mathbf{y} + \mu^2\|\mathbf{y}\|^2$.

Inequalities Involving the Norm

Theorem

Let $\mathbf{x}, \mathbf{y} \in \mathbb{R}^n$. Then:

- (i) $|\mathbf{x} \cdot \mathbf{y}| \leq \|\mathbf{x}\| \|\mathbf{y}\|$ (**Cauchy-Schwarz Inequality**);
- (ii) $\|\mathbf{x} + \mathbf{y}\| \leq \|\mathbf{x}\| + \|\mathbf{y}\|$ (**Triangle Inequality**);
- (iii) $|\|\mathbf{x}\| - \|\mathbf{y}\|| \leq \|\mathbf{x} - \mathbf{y}\|$;
- (iv) if, for some $\alpha > 0$, $\|\mathbf{x} + \lambda \mathbf{y}\| \geq \|\mathbf{x}\|$ whenever $0 < \lambda < \alpha$, then $\mathbf{x} \cdot \mathbf{y} \geq 0$.

- We only prove (iv), since (i), (ii), and (iii) are standard results.

Let $\alpha > 0$ be such that $\|\mathbf{x} + \lambda \mathbf{y}\| \geq \|\mathbf{x}\|$ whenever $0 < \lambda < \alpha$. Then, whenever $0 < \lambda < \alpha$,

$$\|\mathbf{x}\|^2 \leq \|\mathbf{x} + \lambda \mathbf{y}\|^2 = \|\mathbf{x}\|^2 + 2\lambda \mathbf{x} \cdot \mathbf{y} + \lambda^2 \|\mathbf{y}\|^2.$$

Hence $\mathbf{x} \cdot \mathbf{y} + \frac{1}{2}\lambda \|\mathbf{y}\|^2 \geq 0$. Letting $\lambda \rightarrow 0_+$ in the last inequality, we deduce that $\mathbf{x} \cdot \mathbf{y} \geq 0$.

Angle Between Vectors

- The Cauchy-Schwarz inequality allows us to introduce the concept of angle into \mathbb{R}^n .
- The angle between non-zero vectors \mathbf{x} and \mathbf{y} of \mathbb{R}^n is the unique real number θ satisfying the conditions

$$\cos \theta = \frac{\mathbf{x} \cdot \mathbf{y}}{\|\mathbf{x}\| \|\mathbf{y}\|} \quad \text{and} \quad 0 \leq \theta \leq \pi.$$

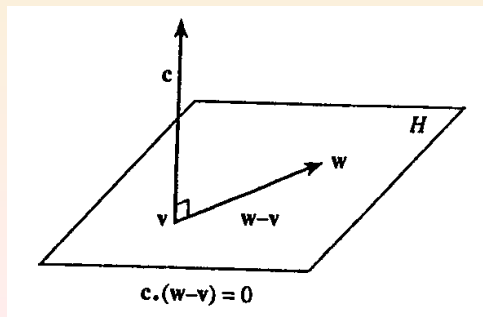
- This definition accords with the usual one of elementary geometry.
- The angle between \mathbf{x} and \mathbf{y} is called **acute** or **obtuse** according as $\mathbf{x} \cdot \mathbf{y}$ is positive or negative.
- Vectors \mathbf{x} and \mathbf{y} , whether zero or not, are said to be **orthogonal** if $\mathbf{x} \cdot \mathbf{y} = 0$.

Normal Vectors to a Hyperplane

- Consider a hyperplane H in \mathbb{R}^n with equation $c_0 + c_1x_1 + \cdots + c_nx_n = 0$.
- This equation can be written in the form $c_0 + \mathbf{c} \cdot \mathbf{x} = 0$, where \mathbf{c} is the non-zero vector (c_1, \dots, c_n) and \mathbf{x} is (x_1, \dots, x_n) .
- Such a vector \mathbf{c} is said to be a **normal vector** to H .
- By the discussion on the representation of hyperplanes by means of linear equations, it follows that the normal vectors of H are precisely those vectors of the form $\lambda \mathbf{c}$ for some non-zero scalar λ .
- Thus H has exactly two unit normal vectors, namely $\pm \frac{\mathbf{c}}{\|\mathbf{c}\|}$.
- Hence, given any hyperplane H in \mathbb{R}^n , it may be assumed that it has an equation of the form $c_0 + \mathbf{c} \cdot \mathbf{x} = 0$, where \mathbf{c} is a unit vector.

Normal Vectors to a Hyperplane (Cont'd)

- This concept of a normal vector generalizes the one familiar in elementary geometry.
- Suppose that \mathbf{v} and \mathbf{w} lie in a hyperplane H in \mathbb{R}^n with equation $c_0 + \mathbf{c} \cdot \mathbf{x} = 0$. Then $c_0 + \mathbf{c} \cdot \mathbf{v} = 0$ and $c_0 + \mathbf{c} \cdot \mathbf{w} = 0$. So $\mathbf{c} \cdot (\mathbf{w} - \mathbf{v}) = 0$.
- This shows that \mathbf{c} is orthogonal to every vector which is the difference of two vectors in H .



Orthogonal Complement

- Let A be a subspace of \mathbb{R}^n .
- Then the **orthogonal complement** A^\perp of A is the set of all those vectors in \mathbb{R}^n which are orthogonal to all the vectors in A , i.e.,

$$A^\perp = \{\mathbf{x} \in \mathbb{R}^n : \mathbf{x} \cdot \mathbf{a} = 0, \text{ for all } \mathbf{a} \in A\}.$$

- It follows easily from this definition that A^\perp is a subspace of \mathbb{R}^n which intersects A in the set $\{\mathbf{0}\}$.
- A standard result of linear algebra asserts that each vector of \mathbb{R}^n can be expressed uniquely in the form $\mathbf{a} + \mathbf{b}$, where $\mathbf{a} \in A$ and $\mathbf{b} \in A^\perp$.
- Thus $A + A^\perp = \mathbb{R}^n$.

Orthonormal Sequences

- A sequence $\mathbf{u}_1, \dots, \mathbf{u}_m$ of vectors in \mathbb{R}^n is said to be an **orthonormal sequence** if $\mathbf{u}_i \cdot \mathbf{u}_j$ is 1 or 0 according as $i = j$ or $i \neq j$.
- The simplest example of such a sequence is the sequence $\mathbf{e}_1, \dots, \mathbf{e}_n$ of elementary vectors in \mathbb{R}^n .
- In an orthonormal sequence, each term is a unit vector, each two terms are orthogonal, and no two terms are the same.
- The terms of an orthonormal sequence $\mathbf{u}_1, \dots, \mathbf{u}_m$ in \mathbb{R}^n form a linearly independent set $\{\mathbf{u}_1, \dots, \mathbf{u}_m\}$.

To see this, suppose that scalars $\lambda_1, \dots, \lambda_m$ are such that $\lambda_1 \mathbf{u}_1 + \dots + \lambda_m \mathbf{u}_m = \mathbf{0}$. Then, for $i = 1, \dots, m$,

$$\lambda_i = (\lambda_1 \mathbf{u}_1 + \dots + \lambda_m \mathbf{u}_m) \cdot \mathbf{u}_i = \mathbf{0} \cdot \mathbf{u}_i = 0.$$

This shows that $\{\mathbf{u}_1, \dots, \mathbf{u}_m\}$ is linearly independent.

- Hence $\{\mathbf{u}_1, \dots, \mathbf{u}_m\}$ is an **orthonormal basis** for the subspace $\text{lin}\{\mathbf{u}_1, \dots, \mathbf{u}_m\}$ of \mathbb{R}^n .

Orthonormal Sequences (Cont'd)

- Thus each point \mathbf{x} of $\text{lin}\{\mathbf{u}_1, \dots, \mathbf{u}_m\}$ can be written uniquely as a linear combination of $\mathbf{u}_1, \dots, \mathbf{u}_m$, say

$$\mathbf{x} = \mu_1 \mathbf{u}_1 + \cdots + \mu_m \mathbf{u}_m.$$

Then, for $i = 1, \dots, m$,

$$\mathbf{x} \cdot \mathbf{u}_i = (\mu_1 \mathbf{u}_1 + \cdots + \mu_m \mathbf{u}_m) \cdot \mathbf{u}_i = \mu_i.$$

We conclude that

$$\mathbf{x} = (\mathbf{x} \cdot \mathbf{u}_1) \mathbf{u}_1 + \cdots + (\mathbf{x} \cdot \mathbf{u}_m) \mathbf{u}_m.$$

Congruences in \mathbb{R}^2

- A congruence transformation in elementary plane geometry is a transformation of the plane which preserves distance.
- Examples of such transformations are reflections, rotations, translations, and combinations of these.
- Algebraically, the congruence transformations of \mathbb{R}^2 are precisely those affine transformations $T : \mathbb{R}^2 \rightarrow \mathbb{R}^2$ that can be expressed in the form

$$T(\mathbf{x}) = \mathbf{Q}\mathbf{x} + \mathbf{q},$$

where \mathbf{Q} is a 2×2 orthogonal matrix and \mathbf{q} is a 2×1 matrix.

Congruence Transformations

- A mapping $T : \mathbb{R}^n \rightarrow \mathbb{R}^n$ is said to be a **congruence transformation** of \mathbb{R}^n if

$$\|T(\mathbf{x}) - T(\mathbf{y})\| = \|\mathbf{x} - \mathbf{y}\|, \quad \text{for all } \mathbf{x}, \mathbf{y} \in \mathbb{R}^n,$$

i.e., T preserves distance.

- We use a superscript T to denote the transpose of a matrix or a vector.
- Thus, recalling that we identify a point $\mathbf{x} = (x_1, \dots, x_n)$ of \mathbb{R}^n with a column vector in the natural way, we see that $\mathbf{x}^T \mathbf{x}$ is the 1×1 matrix whose single element is the scalar $x_1^2 + \dots + x_n^2$.
- We identify this scalar with the matrix $\mathbf{x}^T \mathbf{x}$ itself, so that we may write

$$\|\mathbf{x}\|^2 = x_1^2 + \dots + x_n^2 = \mathbf{x}^T \mathbf{x}.$$

Affine Transformations and Congruences

- We now show that an affine transformation $T : \mathbb{R}^n \rightarrow \mathbb{R}^n$, which is defined by an equation of the form $T(\mathbf{x}) = \mathbf{Q}\mathbf{x} + \mathbf{q}$, where \mathbf{Q} is an $n \times n$ orthogonal matrix and \mathbf{q} is an $n \times 1$ matrix, is a congruence transformation of \mathbb{R}^n .
- Let $\mathbf{x}, \mathbf{y} \in \mathbb{R}^n$. Then

$$\begin{aligned}\|T(\mathbf{x}) - T(\mathbf{y})\|^2 &= \|\mathbf{Q}(\mathbf{x} - \mathbf{y})\|^2 \\ &= (\mathbf{Q}(\mathbf{x} - \mathbf{y}))^\top (\mathbf{Q}(\mathbf{x} - \mathbf{y})) \\ &= (\mathbf{x} - \mathbf{y})^\top \mathbf{Q}^\top \mathbf{Q} (\mathbf{x} - \mathbf{y}) \\ &= (\mathbf{x} - \mathbf{y})^\top (\mathbf{x} - \mathbf{y}) \\ &= \|\mathbf{x} - \mathbf{y}\|^2.\end{aligned}$$

Hence $\|T(\mathbf{x}) - T(\mathbf{y})\| = \|\mathbf{x} - \mathbf{y}\|$. This shows that T is a congruence transformation of \mathbb{R}^n .

Congruences and Affine Transformations

Theorem

Let $T : \mathbb{R}^n \rightarrow \mathbb{R}^n$ be a congruence transformation of \mathbb{R}^n . Then there exist an $n \times n$ orthogonal matrix \mathbf{Q} and an $n \times 1$ matrix \mathbf{q} such that $T(\mathbf{x}) = \mathbf{Q}\mathbf{x} + \mathbf{q}$, for all \mathbf{x} in \mathbb{R}^n .

- Let $\mathbf{x}, \mathbf{y} \in \mathbb{R}^n$. Define a mapping $f : \mathbb{R}^n \rightarrow \mathbb{R}^n$ by the equation $f(\mathbf{x}) = T(\mathbf{x}) - T(\mathbf{0})$. Since T preserves distance,

$$\|f(\mathbf{x})\| = \|T(\mathbf{x}) - T(\mathbf{0})\| = \|\mathbf{x} - \mathbf{0}\| = \|\mathbf{x}\|.$$

So f preserves norms.

Also

$$\|f(\mathbf{x}) - f(\mathbf{y})\|^2 = \|T(\mathbf{x}) - T(\mathbf{y})\|^2 = \|\mathbf{x} - \mathbf{y}\|^2.$$

So

$$\|f(\mathbf{x})\|^2 - 2f(\mathbf{x}) \cdot f(\mathbf{y}) + \|f(\mathbf{y})\|^2 = \|\mathbf{x}\|^2 - 2\mathbf{x} \cdot \mathbf{y} + \|\mathbf{y}\|^2.$$

Congruences and Affine Transformations (Cont'd)

- Since $\|f(\mathbf{x})\| = \|\mathbf{x}\|$ and $\|f(\mathbf{y})\| = \|\mathbf{y}\|$, we can deduce from the last equation that $f(\mathbf{x}) \cdot f(\mathbf{y}) = \mathbf{x} \cdot \mathbf{y}$.

Thus, f preserves inner products.

It follows that $f(\mathbf{e}_1), \dots, f(\mathbf{e}_n)$ is an orthonormal sequence in \mathbb{R}^n .

Hence

$$f(\mathbf{x}) = (f(\mathbf{x}) \cdot f(\mathbf{e}_1))f(\mathbf{e}_1) + \cdots + (f(\mathbf{x}) \cdot f(\mathbf{e}_n))f(\mathbf{e}_n).$$

Writing \mathbf{x} for (x_1, \dots, x_n) and \mathbf{Q} for the $n \times n$ orthogonal matrix whose columns are $f(\mathbf{e}_1), \dots, f(\mathbf{e}_n)$, we deduce that

$$\begin{aligned} f(\mathbf{x}) &= (\mathbf{x} \cdot \mathbf{e}_1)f(\mathbf{e}_1) + \cdots + (\mathbf{x} \cdot \mathbf{e}_n)f(\mathbf{e}_n) \\ &= x_1 f(\mathbf{e}_1) + \cdots + x_n f(\mathbf{e}_n) \\ &= \mathbf{Q}\mathbf{x}. \end{aligned}$$

The proof is completed by putting $\mathbf{q} = T(\mathbf{0})$.

Congruent Subsets

- We have thus identified the congruence transformations $T : \mathbb{R}^n \rightarrow \mathbb{R}^n$ of \mathbb{R}^n as being precisely those affine transformations $T : \mathbb{R}^n \rightarrow \mathbb{R}^n$ which can be expressed in the form $T(\mathbf{x}) = \mathbf{Q}\mathbf{x} + \mathbf{q}$, where \mathbf{Q} is an $n \times n$ orthogonal matrix and \mathbf{q} is an $n \times 1$ matrix.
- Sets A and B in \mathbb{R}^n are said to be **congruent** if there is a congruence transformation T of \mathbb{R}^n such that $T(A) = B$.
- It is easy to verify that congruence is an equivalence relation on the family of all subsets of \mathbb{R}^n .
- In elementary geometry, any two points are congruent, any two lines are congruent, and any two planes are congruent.

Congruent Flats

Theorem

Let A and B be r -flats in \mathbb{R}^n . Then A and B are congruent.

- We consider the non-trivial cases when $r \geq 1$.

First we show that the r -flat A is congruent to the r -flat R_r defined by the equation

$$R_r = \{(x_1, \dots, x_r, 0, \dots, 0) : x_1, \dots, x_r \in \mathbb{R}\}.$$

Let $\mathbf{a} \in A$. Then $A - \mathbf{a}$ is an r -dimensional subspace of \mathbb{R}^n .

Let $\{\mathbf{u}_1, \dots, \mathbf{u}_n\}$ be an orthonormal basis for \mathbb{R}^n such that $\{\mathbf{u}_1, \dots, \mathbf{u}_r\}$ is an orthonormal basis for $A - \mathbf{a}$. Define a congruence transformation T of \mathbb{R}^n by the equation

$$T(\mathbf{x}) = [\mathbf{u}_1, \dots, \mathbf{u}_n]\mathbf{x} + \mathbf{a}.$$

Then $T(R_r) = A$. So A and R_r are congruent. Similarly, B and R_r are congruent. Thus A and B are congruent.

Congruent Copies of a Set

- We now show how, given any r -dimensional set A in \mathbb{R}^n with $1 \leq r \leq n$, it is possible to find a congruent copy of A in the space \mathbb{R}^r .
- Moreover, we show that any two such congruent copies of A in \mathbb{R}^r are themselves congruent to one another in \mathbb{R}^r .
- Let A be an r -dimensional ($1 \leq r \leq n$) set in \mathbb{R}^n . Then $\text{aff}A$ is an r -flat. So by the theorem, there is a congruence transformation of \mathbb{R}^n which maps $\text{aff}A$ onto the r -flat

$$R_r = \{(x_1, \dots, x_r, 0, \dots, 0) : x_1, \dots, x_r \in \mathbb{R}\}.$$

It follows that there is a set B in R_r , which is congruent to A .

Let $i: R_r \rightarrow \mathbb{R}^r$ be the mapping that identifies each point $(x_1, \dots, x_r, 0, \dots, 0)$ of R_r with the point (x_1, \dots, x_r) of \mathbb{R}^r . Then $i(B)$ is a set lying in \mathbb{R}^r which is a congruent copy of the set A in \mathbb{R}^n .

- In general, there will be an infinite number of such copies. We now see how any two of these copies of A are related.

Congruent Copies of a Set (Cont'd)

- Let $i(B)$ and $i(C)$ be congruent copies of A in \mathbb{R}^r , where B and C are congruent to A in \mathbb{R}^n and lie in R_r .

Then there is a congruence transformation T of \mathbb{R}^n such that $T(B) = C$, and which maps R_r onto itself.

By considering the images of $\mathbf{0}$ and the elementary vectors $\mathbf{e}_1, \dots, \mathbf{e}_r$ under T , it follows that T can be expressed in the form

$$T(\mathbf{x}) = \begin{bmatrix} \mathbf{Q} & \mathbf{0} \\ \mathbf{0} & * \end{bmatrix} \mathbf{x} + \begin{bmatrix} \mathbf{q} \\ \mathbf{0} \end{bmatrix},$$

where \mathbf{Q} is an $r \times r$ orthogonal matrix, \mathbf{q} is an $r \times 1$ matrix, and $\mathbf{0}$ represents zero matrices of suitable shapes and sizes.

Denote by T_r the congruence transformation of \mathbb{R}^r defined by the equation $T_r(\mathbf{x}) = \mathbf{Q}\mathbf{x} + \mathbf{q}$, where $\mathbf{x} = (x_1, \dots, x_r)$.

Then $T_r(i(B)) = i(C)$. This shows that the congruent copies $i(B)$ and $i(C)$ of A in \mathbb{R}^r are congruent to one another in \mathbb{R}^r .

Subsection 7

Open Sets and Closed Sets

Open and Closed Balls

- Let $\mathbf{a} \in \mathbb{R}^n$ and $r > 0$.
- Then the **open ball** $B(\mathbf{a}; r)$ (**closed ball** $B[\mathbf{a}; r]$) with **center** \mathbf{a} and **radius** r is the set of all points of \mathbb{R}^n whose distance from \mathbf{a} is less than (less than or equal to) r , i.e.,

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

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

- In \mathbb{R}^1 the open (closed) ball with center a and radius r is the open (closed) **interval** $(a - r, a + r)$ ($[a - r, a + r]$).
- In \mathbb{R}^2 open (closed) balls are referred to as **open (closed) discs**.

Open and Closed Unit Balls

- The balls $B(\mathbf{0}; 1)$ and $B[\mathbf{0}; 1]$ in \mathbb{R}^n are called, respectively, the **open unit ball** and the **closed unit ball**.
- If we denote them, respectively, by V and U , then

$$V = \{\mathbf{x} \in \mathbb{R}^n : \|\mathbf{x}\| < 1\} \quad \text{and} \quad U = \{\mathbf{x} \in \mathbb{R}^n : \|\mathbf{x}\| \leq 1\}.$$

- It follows that $B(\mathbf{a}; r) = \mathbf{a} + rV$ and $B[\mathbf{a}; r] = \mathbf{a} + rU$.
- We adopt U as the standard notation for the closed unit ball.

Open Sets

- A point \mathbf{a} of a set A in \mathbb{R}^n is said to be an **interior point** of A if it is the center of some open ball which lies in A , i.e. if there exists some $r > 0$ such that $B(\mathbf{a}; r) \subseteq A$.
- The set of interior points of A is called the **interior** of A and is denoted by $\text{int}A$.
- Clearly $\text{int}B \subseteq \text{int}A$ when $B \subseteq A$.
- A set in \mathbb{R}^n , each of whose points is an interior point of the set, is said to be **open**.
- Since $\text{int}A \subseteq A$ is always true, A is open if and only if $\text{int}A = A$.
- Clearly the sets \emptyset and \mathbb{R}^n are open.

Balls, Halfspaces, Hyperplanes

Theorem

In \mathbb{R}^n open balls and open halfspaces are open, and hyperplanes have empty interiors.

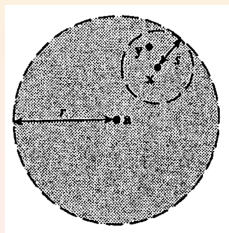
- Consider the open ball $B(\mathbf{a}; r)$, where $\mathbf{a} \in \mathbb{R}^n$ and $r > 0$. Let $\mathbf{x} \in B(\mathbf{a}; r)$. We prove that $B(\mathbf{a}; r)$ is open by showing that $B(\mathbf{x}; s) \subseteq B(\mathbf{a}; r)$, where s is the positive number $r - \|\mathbf{x} - \mathbf{a}\|$.

Let $\mathbf{y} \in B(\mathbf{x}; s)$. Then $\|\mathbf{y} - \mathbf{x}\| < s$.

So by the triangle inequality

$$\begin{aligned} \|\mathbf{y} - \mathbf{a}\| &\leq \|\mathbf{y} - \mathbf{x} + \mathbf{x} - \mathbf{a}\| \\ &\leq \|\mathbf{y} - \mathbf{x}\| + \|\mathbf{x} - \mathbf{a}\| \\ &< s + \|\mathbf{x} - \mathbf{a}\| = r. \end{aligned}$$

Thus $\mathbf{y} \in B(\mathbf{a}; r)$. So $B(\mathbf{x}; s) \subseteq B(\mathbf{a}; r)$.



Balls, Halfspaces, Hyperplanes (Cont'd)

- Consider the open halfspace A in \mathbb{R}^n which is defined by the inequality $c_0 + \mathbf{c} \cdot \mathbf{x} > 0$, where \mathbf{c} is a unit vector. Let $\mathbf{a} \in A$. We prove that A is open by showing that $B(\mathbf{a}; r) \subseteq A$, where r is the positive number $c_0 + \mathbf{c} \cdot \mathbf{a}$. Let $\mathbf{y} \in B(\mathbf{a}; r)$. Then $\|\mathbf{y} - \mathbf{a}\| < r$. Moreover,

$$c_0 + \mathbf{c} \cdot \mathbf{y} = c_0 + \mathbf{c} \cdot \mathbf{a} + \mathbf{c} \cdot (\mathbf{y} - \mathbf{a}) = r + \mathbf{c} \cdot (\mathbf{y} - \mathbf{a}) > 0,$$

since, by the Cauchy-Schwarz Inequality, $|\mathbf{c} \cdot (\mathbf{y} - \mathbf{a})| \leq \|\mathbf{y} - \mathbf{a}\| < r$. Thus $\mathbf{y} \in A$. So $B(\mathbf{a}; r) \subseteq A$.

- Consider the hyperplane H in \mathbb{R}^n with equation $c_0 + \mathbf{c} \cdot \mathbf{x} = 0$, where \mathbf{c} is a unit vector. We show that no point \mathbf{a} of H is an interior point of H . Let $r > 0$. Then $\mathbf{a} + \frac{1}{2}r\mathbf{c} \notin H$ and $\|\mathbf{a} + \frac{1}{2}r\mathbf{c} - \mathbf{a}\| = \frac{1}{2}r$. Therefore, $\mathbf{a} + \frac{1}{2}r\mathbf{c} \in B(\mathbf{a}; r)$ and $B(\mathbf{a}; r) \not\subseteq H$. Hence, \mathbf{a} is not an interior point of H . So H has an empty interior.

Properties of the Interior

Corollary

Let A be a set in \mathbb{R}^n . Then $\text{int}A$ is open and $\text{int}(\text{int}A) = \text{int}A$.

- If $\mathbf{a} \in \text{int}A$, then there exists $r > 0$ such that $B(\mathbf{a}; r) \subseteq A$. Since $B(\mathbf{a}; r)$ is open,

$$B(\mathbf{a}; r) = \text{int}(B(\mathbf{a}; r)) \subseteq \text{int}A.$$

Hence, $\mathbf{a} \in \text{int}(\text{int}A)$. So $\text{int}A \subseteq \text{int}(\text{int}A)$. Thus, $\text{int}A$ is open and $\text{int}(\text{int}A) = \text{int}A$.

Properties of Open Sets

Theorem

In \mathbb{R}^n every union and every finite intersection of open sets is open.

- Let A be the union of a family $(A_i : i \in I)$ of open sets in \mathbb{R}^n . If $\mathbf{a} \in A$, then $\mathbf{a} \in A_i$, for some $i \in I$. Since A_i is open, there is an $r > 0$ such that $B(\mathbf{a}; r) \subseteq A_i$. Hence, $B(\mathbf{a}; r) \subseteq A$. Thus, A is open.
- Let A be the intersection of the open sets A_1, \dots, A_m in \mathbb{R}^n . If $\mathbf{a} \in A$, then $\mathbf{a} \in A_1, \dots, \mathbf{a} \in A_m$. Since A_1, \dots, A_m are open, there exist $r_1, \dots, r_m > 0$ such that $B(\mathbf{a}; r_1) \subseteq A_1, \dots, B(\mathbf{a}; r_m) \subseteq A_m$. Let $r = \min\{r_1, \dots, r_m\}$. Then $r > 0$ and

$$B(\mathbf{a}; r) \subseteq B(\mathbf{a}; r_1) \cap \dots \cap B(\mathbf{a}; r_m) \subseteq A_1 \cap \dots \cap A_m = A.$$

Thus A is open.

Intersections of Open Sets

- An arbitrary intersection of open sets in \mathbb{R}^n need not be open.
- To see this, we note that the intersection of the sequence

$$V, \frac{1}{2}V, \frac{1}{3}V, \dots, \frac{1}{k}V, \dots$$

of open balls centered at the origin of \mathbb{R}^n is the singleton set $\{\mathbf{0}\}$, which is not open.

Closure of a Set

- A point \mathbf{a} of \mathbb{R}^n is said to be a **closure point** of a set A in \mathbb{R}^n if every open ball with center \mathbf{a} meets A , i.e., if for every $r > 0$ the ball $B(\mathbf{a}; r)$ meets A .
- The set of closure points of A is called the **closure** of A and is denoted by $\text{cl}A$.
- Clearly $A \subseteq \text{cl}A$.
- Also $\text{cl}B \subseteq \text{cl}A$ whenever $B \subseteq A$.
- Roughly speaking, the closure of A is the set of all points in \mathbb{R}^n which either lie in A or are arbitrarily close to A .
- Thus, in \mathbb{R}^1 the closures of the intervals $(0, 1]$, $(0, 1)$, $[0, 1)$ are all equal to the interval $[0, 1]$.
- In \mathbb{R}^2 the closures of the discs $B(\mathbf{a}; r)$ and $B[\mathbf{a}; r]$ are both equal to the disc $B[\mathbf{a}; r]$.

Closed Sets

- A set in \mathbb{R}^n each of whose closure points lies in the set is said to be **closed**.
- Thus a set A in \mathbb{R}^n is closed if and only if $\text{cl}A \subseteq A$.
- Since $A \subseteq \text{cl}A$ is always true, A is closed if and only if $\text{cl}A = A$.
- Clearly the sets \emptyset and \mathbb{R}^n are closed.
- Thus the sets \emptyset and \mathbb{R}^n are both open and closed.
- It can be shown that they are the only sets in \mathbb{R}^n with this property.
- A set in \mathbb{R}^n may be neither open nor closed.
- For example, in \mathbb{R}^1 the interval $[0,1)$ is such a set.

Closure and Interior

- For each set A in \mathbb{R}^n , we denote by A^c the complement of A in \mathbb{R}^n , i.e., the set $\mathbb{R}^n \setminus A$.

Theorem

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

- If $\mathbf{x} \in \text{cl}A$, then each open ball with center \mathbf{x} contains a point of A . So \mathbf{x} cannot belong to $\text{int}A^c$, i.e., $\mathbf{x} \in (\text{int}A^c)^c$.

If $\mathbf{x} \in (\text{int}A^c)^c$, then each open ball with center \mathbf{x} must contain a point of A , i.e., $\mathbf{x} \in \text{cl}A$.

Thus $\text{cl}A = (\text{int}A^c)^c$.

Closed and Open Sets

Theorem

A set in \mathbb{R}^n is closed if and only if its complement in \mathbb{R}^n is open.

- Let A be a set in \mathbb{R}^n . Suppose first that A is closed. Then $\text{cl}A = A$. It follows from a previous corollary and the preceding theorem that A^c is the open set $\text{int}A^c$. Suppose next that A^c is open. Then $\text{int}A^c = A^c$. It follows from the theorem that $\text{cl}A = A$, i.e., A is closed.

Corollary

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

- Now $\text{int}A^c$ is open by a previous corollary. Hence by the theorem its complement $\text{cl}A$ is closed.

Properties of Closed Sets

Theorem

In \mathbb{R}^n every intersection and every finite union of closed sets is closed.

- Let $(A_i : i \in I)$ be a family of closed sets in \mathbb{R}^n . Then, for each $i \in I$, A_i^c is open. By a previous theorem, $\bigcup(A_i^c : i \in I)$ is open. Hence

$$\bigcap(A_i : i \in I) = \left(\bigcup(A_i^c : i \in I)\right)^c$$

is closed.

Now let A_1, \dots, A_m be closed sets in \mathbb{R}^n . Then A_1^c, \dots, A_m^c are open. By a previous theorem, $A_1^c \cap \dots \cap A_m^c$ is open. Hence

$$A_1 \cup \dots \cup A_m = \left(A_1^c \cap \dots \cap A_m^c\right)^c$$

is closed.

Closures and Unions

Corollary

Let A_1, \dots, A_m be sets in \mathbb{R}^n . Then

$$\text{cl}(A_1 \cup \dots \cup A_m) = \text{cl}A_1 \cup \dots \cup \text{cl}A_m.$$

- Since $A_1 \cup \dots \cup A_m$ is contained in the closed set $\text{cl}A_1 \cup \dots \cup \text{cl}A_m$,

$$\text{cl}(A_1 \cup \dots \cup A_m) \subseteq \text{cl}A_1 \cup \dots \cup \text{cl}A_m.$$

Trivially,

$$\text{cl}(A_1 \cup \dots \cup A_m) \supseteq \text{cl}A_1 \cup \dots \cup \text{cl}A_m.$$

Thus,

$$\text{cl}(A_1 \cup \dots \cup A_m) = \text{cl}A_1 \cup \dots \cup \text{cl}A_m.$$

Closed Balls, Closed Halfspaces, Flats

Theorem

In \mathbb{R}^n closed balls, closed halfspaces and flats are closed.

- Let A be the closed ball $B[\mathbf{a}; r]$, where $\mathbf{a} \in \mathbb{R}^n$ and $r > 0$. We prove that A^c is open. Let $\mathbf{x} \in A^c$. Then we show that $B(\mathbf{x}; s) \subseteq A^c$, where s is the positive number $\|\mathbf{x} - \mathbf{a}\| - r$. Suppose that this is not the case. Then there is some point of A , \mathbf{y} say, which lies in $B(\mathbf{x}; s)$. Now

$$\|\mathbf{x} - \mathbf{a}\| = \|\mathbf{x} - \mathbf{y} + \mathbf{y} - \mathbf{a}\| < s + r = \|\mathbf{x} - \mathbf{a}\|,$$

which is impossible. Hence $B(\mathbf{x}; s) \subseteq A^c$.

A previous theorem shows that open halfspaces in \mathbb{R}^n are open. Hence their complements in \mathbb{R}^n , i.e., the closed halfspaces, are closed.

In \mathbb{R}^n each hyperplane is the intersection of two closed halfspaces. So it is closed. By a previous corollary, each flat in \mathbb{R}^n is an intersection of hyperplanes. So it is closed.

Boundaries

- A point \mathbf{a} of \mathbb{R}^n is said to be a **boundary point** of a set A in \mathbb{R}^n if every open ball with center \mathbf{a} meets both A and its complement A^c .
- The set of boundary points of A is called the **boundary** of A and is denoted by $\text{bd}A$.
- Thus a boundary point of a set in \mathbb{R}^n is a point of \mathbb{R}^n which is arbitrarily close both to the set and its complement.
- It follows from the preceding definitions that $\text{bd}A = (\text{cl}A) \cap (\text{cl}A^c)$.
- Hence the boundary of a set in \mathbb{R}^n is always closed, being the intersection of two closed sets.

More Properties of Boundaries

- A boundary point of a set in \mathbb{R}^n may or may not belong to the set itself.
- For example, in \mathbb{R}^1 the interval $[0,1)$ contains its boundary point 0, but not its boundary point 1.
- For any set A in \mathbb{R}^n , the sets A and A^c have the same boundary.
- Moreover, the sets $\text{int}A$, $\text{bd}A$, $\text{int}A^c$ form a partition of \mathbb{R}^n .
- Open (closed) sets in \mathbb{R}^n are characterized by the property that they contain none (all) of their boundary points.

Dependence on Ambient Space

- The above definitions of the interior and the boundary of a set depend upon the space in which the set is embedded.
- For example, a closed line segment in \mathbb{R}^2 has an empty interior and is its own boundary.
- The same line segment considered as a subset of \mathbb{R}^1 has for its interior the set of all of its points with the exception of its two boundary points, these forming its boundary in \mathbb{R}^1 .
- The latter interior and boundary, obtained by regarding the one-dimensional line segment as a set in the one-dimensional space \mathbb{R}^1 , correspond to what may be thought of as the “intrinsic” interior and boundary of the segment.

Relative Interior

- A point \mathbf{a} of a set A in \mathbb{R}^n is said to be a **relative interior point** of A if it is the center of some open ball whose intersection with $\text{aff}A$ is contained in A , i.e., if there exists $r > 0$ such that

$$B(\mathbf{a}; r) \cap \text{aff}A \subseteq A.$$

- The set of all relative interior points of A is called the **relative interior** of A and is denoted by $\text{ri}A$.
- The relative interior of an n -dimensional set in \mathbb{R}^n coincides with its interior.
- The relative interior of any flat in \mathbb{R}^n is itself.

Relative Boundary

- A point \mathbf{a} of \mathbb{R}^n is said to be a **relative boundary point** of a set A in \mathbb{R}^n if it lies in the closure of A but not in its relative interior.
- The set of all relative boundary points of A is called the **relative boundary** of A and is denoted by $\text{rebd}A$.
- The relative boundary of an n -dimensional set in \mathbb{R}^n coincides with its boundary.

Properties of Relative Interior

- It is to be noted that while the inclusion $B \subseteq A$ implies both $\text{int}B \subseteq \text{int}A$ and $\text{cl}B \subseteq \text{cl}A$, it does not in general imply $\text{ri}B \subseteq \text{ri}A$.
- For example, if B is one side of a square A in \mathbb{R}^2 , then $\text{ri}B$ and $\text{ri}A$ are non-empty but disjoint.
- If, however, $B \subseteq A$ and $\dim B = \dim A$ or, equivalently, $\text{aff}B = \text{aff}A$, then $\text{ri}B \subseteq \text{ri}A$.

Flats and Relative Boundaries

- Suppose that \mathbf{a} is a point of a set A in \mathbb{R}^n and that \mathbf{x} is a point of $\text{aff}A$ not lying in A .
- Define a scalar λ_0 by the equation

$$\lambda_0 = \sup \{ \lambda \in [0, 1] : (1 - \lambda)\mathbf{a} + \lambda\mathbf{x} \in A \}.$$

- Then $(1 - \lambda_0)\mathbf{a} + \lambda_0\mathbf{x}$ is a relative boundary point of A lying between \mathbf{a} and \mathbf{x} .
- It follows that flats are the only sets in \mathbb{R}^n which have an empty relative boundary.

Subsection 8

Convergence and Compactness

Convergence of Sequences

- In \mathbb{R}^n a sequence $\mathbf{x}_1, \dots, \mathbf{x}_k, \dots$ of points is said to **converge** to a point \mathbf{x} if $\|\mathbf{x}_k - \mathbf{x}\| \rightarrow 0$ as $k \rightarrow \infty$, i.e., if the distance $\|\mathbf{x}_k - \mathbf{x}\|$ between \mathbf{x}_k and \mathbf{x} tends to zero as k tends to infinity.
- We indicate such convergence by writing $\mathbf{x}_k \rightarrow \mathbf{x}$ as $k \rightarrow \infty$, or simply $\mathbf{x}_k \rightarrow \mathbf{x}$.
- This convergence for sequences of points in \mathbb{R}^n coincides with that of classical convergence for real sequences.

Properties of Convergence

- The inequality $|\|\mathbf{x}\| - \|\mathbf{y}\|| \leq \|\mathbf{x} - \mathbf{y}\|$ proven previously, shows that

$$|\|\mathbf{x}_k\| - \|\mathbf{x}\|| \leq \|\mathbf{x}_k - \mathbf{x}\|.$$

Hence $\|\mathbf{x}_k\| \rightarrow \|\mathbf{x}\|$ as $k \rightarrow \infty$ whenever $\mathbf{x}_k \rightarrow \mathbf{x}$ as $k \rightarrow \infty$.

- The triangle inequality shows that

$$\|\mathbf{x}_i - \mathbf{x}_j\| \leq \|\mathbf{x}_i - \mathbf{x}\| + \|\mathbf{x} - \mathbf{x}_j\|.$$

Hence $\|\mathbf{x}_i - \mathbf{x}_j\| \rightarrow 0$ as $i, j \rightarrow \infty$ whenever $\mathbf{x}_k \rightarrow \mathbf{x}$ as $k \rightarrow \infty$.

Convergence and Coordinate-wise Convergence

- Suppose that $\mathbf{x}_k = (x_{k1}, \dots, x_{kn})$ for $k = 1, 2, \dots$ and that $\mathbf{x} = (x_1, \dots, x_n)$.
- Then, for $i = 1, \dots, n$, we have

$$|x_{ki} - x_i|^2 \leq (x_{k1} - x_1)^2 + \dots + (x_{kn} - x_n)^2 = \|\mathbf{x}_k - \mathbf{x}\|^2.$$

- We also have

$$\begin{aligned}\|\mathbf{x}_k - \mathbf{x}\|^2 &= (x_{k1} - x_1)^2 + \dots + (x_{kn} - x_n)^2 \\ &\leq (|x_{k1} - x_1| + \dots + |x_{kn} - x_n|)^2.\end{aligned}$$

- Hence

$$|x_{ki} - x_i| \leq \|\mathbf{x}_k - \mathbf{x}\| \leq |x_{k1} - x_1| + \dots + |x_{kn} - x_n|.$$

- Thus, $\mathbf{x}_k \rightarrow \mathbf{x}$ if and only if $x_{ki} \rightarrow x_i$, for $i = 1, \dots, n$.
- So the convergence of $\mathbf{x}_1, \dots, \mathbf{x}_k, \dots$ to (x_1, \dots, x_n) is equivalent to the convergence of each of the coordinate sequences $x_{1i}, \dots, x_{ki}, \dots$ for $i = 1, \dots, n$.

Uniqueness and Linearity Properties

- A consequence of coordinate-wise convergence is that a sequence of points in \mathbb{R}^n can converge to at most one point.
- Moreover, if $\mathbf{x}_k \rightarrow \mathbf{x}$, $\mathbf{y}_k \rightarrow \mathbf{y}$ in \mathbb{R}^n and $\lambda_k \rightarrow \lambda$, $\mu_k \rightarrow \mu$ in \mathbb{R} , then

$$\begin{aligned}\mathbf{x}_k \cdot \mathbf{y}_k &\rightarrow \mathbf{x} \cdot \mathbf{y} \quad \text{in } \mathbb{R}; \\ \lambda_k \mathbf{x}_k + \mu_k \mathbf{y}_k &\rightarrow \lambda \mathbf{x} + \mu \mathbf{y} \quad \text{in } \mathbb{R}^n.\end{aligned}$$

Boundedness

- We recall that a sequence x_1, \dots, x_k, \dots of real numbers is said to be **bounded** if there exists a real number r such that $|x_k| \leq r$ for $k = 1, 2, \dots$
- Similarly, a sequence $\mathbf{x}_1, \dots, \mathbf{x}_k, \dots$ of points in \mathbb{R}^n is defined to be **bounded** if there exists a real number r such that $\|\mathbf{x}_k\| \leq r$ for $k = 1, 2, \dots$
- Every convergent sequence of real numbers is bounded, and the same is also true for convergent sequences of points in \mathbb{R}^n .

To see this, suppose that $\mathbf{x}_k \rightarrow \mathbf{x}$ in \mathbb{R}^n . By what we proved above, $\|\mathbf{x}_k\| \rightarrow \|\mathbf{x}\|$. So there exists a real number r such that $\|\mathbf{x}_k\| \leq r$ for $k = 1, 2, \dots$

Boundedness and Convergence

- The next theorem generalizes to \mathbb{R}^n the classical result that every bounded sequence of real numbers contains a convergent subsequence.

Theorem

Every bounded sequence of points of \mathbb{R}^n contains a convergent subsequence.

- Let $\mathbf{x}_1, \dots, \mathbf{x}_k, \dots$ be a bounded sequence of points in \mathbb{R}^n . Then each of the n coordinate sequences associated with $\mathbf{x}_1, \dots, \mathbf{x}_k, \dots$ is bounded in \mathbb{R} . In particular, the sequence of the first coordinates of $\mathbf{x}_1, \dots, \mathbf{x}_k, \dots$ is a bounded sequence of real numbers. Thus there exists a subsequence of $\mathbf{x}_1, \dots, \mathbf{x}_k, \dots$ such that the sequence of its first coordinates converges. Similarly, there exists a subsequence of this subsequence of $\mathbf{x}_1, \dots, \mathbf{x}_k, \dots$ such that the sequence of its second coordinates converges.

Boundedness and Convergence (Cont'd)

- After performing this subsequence operation n times in all, we arrive at a subsequence of $\mathbf{x}_1, \dots, \mathbf{x}_k, \dots$ each of whose n coordinate sequences converges.

I.e., we have found a convergent subsequence of $\mathbf{x}_1, \dots, \mathbf{x}_k, \dots$

Closure in Terms of Sequences

Theorem

Let A be a set in \mathbb{R}^n . Then $\mathbf{x} \in \text{cl}A$ if and only if there exists a sequence of points of A which converges to \mathbf{x} .

- Suppose first that $\mathbf{x}_1, \dots, \mathbf{x}_k, \dots$ is a sequence of points of A which converges to a point \mathbf{x} of \mathbb{R}^n . Then, for each $r > 0$, there is some point \mathbf{x}_k of the sequence such that $\|\mathbf{x}_k - \mathbf{x}\| < r$. Hence the open ball $B(\mathbf{x}; r)$ meets A . This shows that $\mathbf{x} \in \text{cl}A$.

Suppose next that $\mathbf{x} \in \text{cl}A$. Then, for each positive integer k , the ball $B(\mathbf{x}; \frac{1}{k})$ meets A . Hence there exists $\mathbf{x}_k \in A$ such that $\|\mathbf{x}_k - \mathbf{x}\| < \frac{1}{k}$. It follows that $\mathbf{x}_1, \dots, \mathbf{x}_k, \dots$ converges to \mathbf{x} .

Closed Sets in Terms of Sequences

Corollary

Let A be a set in \mathbb{R}^n . Then A is closed if and only if each convergent sequence of points of A converges to a point of A .

- The corollary follows from the theorem and the fact that A is closed if and only if $A = \text{cl}A$.

Bounded and Compact Subsets

- The set A in \mathbb{R}^n is said to be **bounded** if there exists a real number r such that $\|\mathbf{a}\| \leq r$ for all $\mathbf{a} \in A$.
- Clearly, a set in \mathbb{R}^n is bounded if and only if each sequence of its points is bounded.
- In \mathbb{R}^n balls and finite sets are bounded, whereas r -flats ($r \geq 1$) are not.
- A previous theorem and a corollary, taken together, show that each sequence of points of a closed bounded set in \mathbb{R}^n contains some subsequence which converges to a point of the set.
- A subset of \mathbb{R}^n is said to be **compact**, if each sequence of its points contains some subsequence that converges to a point of the subset.

Characterization of Compact Subsets

Theorem

Let A be a set in \mathbb{R}^n . Then A is compact if and only if it is both closed and bounded.

- We know that closed bounded subsets of \mathbb{R}^n are compact.

Suppose, then, that A is compact. We show first that A is closed. If $\mathbf{x} \in \text{cl}A$, then, by a previous theorem, there is a sequence of points of A which converges to \mathbf{x} . Every subsequence of such a sequence also converges to \mathbf{x} . The compactness of A and the uniqueness of limits show that $\mathbf{x} \in A$. Hence A is closed.

Suppose next that A is not bounded. Then, for each positive integer k , there must exist a point \mathbf{x}_k of A such that $\|\mathbf{x}_k\| > k$. The sequence $\mathbf{x}_1, \dots, \mathbf{x}_k, \dots$ of points of A contains no bounded subsequence, and hence no convergent subsequence, contrary to the hypothesis that A is compact. Hence A is both closed and bounded.

Compactness and Coverings

Theorem

Let A be a non-empty compact set in \mathbb{R}^n and let $r > 0$. Then there exists a finite number of points $\mathbf{a}_1, \dots, \mathbf{a}_m$ of A such that

$$A \subseteq B(\mathbf{a}_1; r) \cup \dots \cup B(\mathbf{a}_m; r).$$

- We argue by contradiction. Suppose that no such finite number of points of A exists. Let $\mathbf{x}_1 \in A$. Then $A \not\subseteq B(\mathbf{x}_1; r)$. Hence there exists a point \mathbf{x}_2 of A such that $\|\mathbf{x}_2 - \mathbf{x}_1\| \geq r$. Now $A \not\subseteq B(\mathbf{x}_1; r) \cup B(\mathbf{x}_2; r)$. Hence there exists a point \mathbf{x}_3 of A such that $\|\mathbf{x}_3 - \mathbf{x}_1\| \geq r$ and $\|\mathbf{x}_3 - \mathbf{x}_2\| \geq r$. Continuing in this way, we produce a sequence $\mathbf{x}_1, \dots, \mathbf{x}_k, \dots$ of points of A with the property that $\|\mathbf{x}_i - \mathbf{x}_j\| \geq r$ whenever $i \neq j$. Clearly such a sequence cannot contain a convergent subsequence. This contradicts the compactness of A .

Balls of Fixed Radius in a Covering

Lemma

Let A be a compact set in \mathbb{R}^n and let $(U_i : i \in I)$ be a family of open sets in \mathbb{R}^n whose union contains A . Then there exists $r > 0$ such that, for each \mathbf{x} in A , the open ball $B(\mathbf{x}; r)$ is contained in some U_j .

- We argue by contradiction. Suppose that no such $r > 0$ exists. Then, for each positive integer k , there is some point \mathbf{a}_k of A such that $B(\mathbf{a}_k; \frac{1}{k})$ is not contained in any U_i . Since A is compact, the sequence $\mathbf{a}_1, \dots, \mathbf{a}_k, \dots$ has a subsequence which converges to a point \mathbf{a} of A . This point \mathbf{a} must belong to one of the U_i 's, U^* say.

Balls of Fixed Radius in a Covering (Cont'd)

- Since U^* is open, there is an $s > 0$ such that $B(\mathbf{a}; 2s) \subseteq U^*$.

Since some subsequence of $\mathbf{a}_1, \dots, \mathbf{a}_k, \dots$ converges to \mathbf{a} , there are infinitely many positive integers k for which $\|\mathbf{a}_k - \mathbf{a}\| < s$.

Choose one of these positive integers, m say, so large that $\frac{1}{m} < s$.

Let $\mathbf{x} \in B(\mathbf{a}_m; \frac{1}{m})$. Then

$$\|\mathbf{x} - \mathbf{a}\| \leq \|\mathbf{x} - \mathbf{a}_m\| + \|\mathbf{a}_m - \mathbf{a}\| < s + s = 2s.$$

So $\mathbf{x} \in B(\mathbf{a}; 2s)$. Thus $B(\mathbf{a}_m; \frac{1}{m}) \subseteq B(\mathbf{a}; 2s) \subseteq U^*$. This contradicts the assumption that $B(\mathbf{a}_m; \frac{1}{m})$ is not contained in any U_i .

Coverings and Finite Subcoverings

Theorem

Let A be a compact set in \mathbb{R}^n and let $(U_i : i \in I)$ be a family of open sets in \mathbb{R}^n whose union contains A . Then there exists a finite subset I^* of I such that the union of the family $(U_i : i \in I^*)$ contains A .

- We may suppose that A is non-empty. By the lemma, there is an $r > 0$ such that, for each \mathbf{x} in A , the open ball $B(\mathbf{x}; r)$ is contained in some U_i . By the preceding theorem, there exist points $\mathbf{a}_1, \dots, \mathbf{a}_m$ in A such that

$$A \subseteq B(\mathbf{a}_1; r) \cup \dots \cup B(\mathbf{a}_m; r).$$

For each $k = 1, \dots, m$, there exists $i_k \in I$ such that $B(\mathbf{a}_k; r) \subseteq U_{i_k}$. We complete the proof by taking I^* to be the set $\{i_1, \dots, i_m\}$.

Intersection of Families of Compact Sets

Corollary

Let $(A_i : i \in I)$ be a family of compact sets in \mathbb{R}^n whose intersection is empty. Then there exists a finite subset I^* of I such that the intersection of the family $(A_i : i \in I^*)$ is empty.

- Let $i_0 \in I$ and let $I_0 = I \setminus \{i_0\}$. Then, since $\bigcap (A_i : i \in I)$ is empty, $A_{i_0} \subseteq \bigcup (A_i^c : i \in I_0)$. By the theorem, which is applicable since the sets A_i^c are open, being the complements of closed sets in \mathbb{R}^n , there is a finite subset I' of I_0 such that $A_{i_0} \subseteq \bigcup (A_i^c : i \in I')$. It follows that, if I^* denotes the finite subset $I' \cup \{i_0\}$ of I , then $\bigcap (A_i : i \in I^*)$ is empty.

Decreasing Sequence of Compact Sets

Corollary

Let A_1, \dots, A_k, \dots be a sequence of non-empty compact sets in \mathbb{R}^n such that $A_1 \supseteq \dots \supseteq A_k \supseteq \dots$. Then the intersection $\bigcap (A_k : k = 1, 2, \dots)$ is non-empty.

- The intersection of any finite number of members of the family is itself a member of the family. So it is non-empty.

Thus, the result follows from the preceding corollary.

Properties of Linear Combinations of Sets

Theorem

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

- (i) open when A is open and $\lambda \neq 0$;
- (ii) closed when A is compact and B is closed;
- (iii) bounded when A and B are bounded;
- (iv) compact when A and B are compact.

- (i) Let A be open and let $\lambda \neq 0$. If $\mathbf{x} \in \lambda A + \mu B$, then $\mathbf{x} = \lambda \mathbf{a} + \mu \mathbf{b}$ for some $\mathbf{a} \in A$ and $\mathbf{b} \in B$. Since A is open, there is an $r > 0$ such that $\mathbf{a} + rV \subseteq A$, where V is the open unit ball $\{\mathbf{x} \in \mathbb{R}^n : \|\mathbf{x}\| < 1\}$. Thus

$$\mathbf{x} + \lambda rV = \lambda \mathbf{a} + \mu \mathbf{b} + \lambda rV = \lambda(\mathbf{a} + rV) + \mu \mathbf{b} \subseteq \lambda A + \mu B.$$

This shows that $B(\mathbf{x}; |\lambda|r) \subseteq \lambda A + \mu B$. Hence $\lambda A + \mu B$ is open.

Properties of Linear Combinations of Sets (Cont'd)

- (ii) Let A be compact and let B be closed. We consider only the non-trivial case $\mu \neq 0$. If $\mathbf{x} \in \text{cl}(\lambda A + \mu B)$, then there exist sequences $\mathbf{a}_1, \dots, \mathbf{a}_k, \dots$ of points of A , and $\mathbf{b}_1, \dots, \mathbf{b}_k, \dots$ of points of B such that $\lambda \mathbf{a}_k + \mu \mathbf{b}_k \rightarrow \mathbf{x}$ as $k \rightarrow \infty$. Since A is compact, there is a subsequence $\mathbf{a}_{i_1}, \dots, \mathbf{a}_{i_k}, \dots$ of $\mathbf{a}_1, \dots, \mathbf{a}_k, \dots$ which converges to some point \mathbf{a} of A . Thus $\lambda \mathbf{a}_{i_k} + \mu \mathbf{b}_{i_k} \rightarrow \mathbf{x}$ and $\mathbf{b}_{i_k} \rightarrow \frac{\mathbf{x} - \lambda \mathbf{a}}{\mu}$ as $k \rightarrow \infty$. But B is closed, and so $\frac{\mathbf{x} - \lambda \mathbf{a}}{\mu} \in B$. Hence $\mathbf{x} \in \lambda A + \mu B$. Thus $\mathbf{x} \in \text{cl}(\lambda A + \mu B)$ implies that $\mathbf{x} \in \lambda A + \mu B$. This shows that $\lambda A + \mu B$ is closed.
- (iii) Let A and B be bounded. Then there exist real numbers r_1 and r_2 such that $\|\mathbf{a}\| \leq r_1$ and $\|\mathbf{b}\| \leq r_2$ whenever $\mathbf{a} \in A$ and $\mathbf{b} \in B$. If $\mathbf{x} \in \lambda A + \mu B$, then $\mathbf{x} = \lambda \mathbf{a} + \mu \mathbf{b}$ for some $\mathbf{a} \in A$ and $\mathbf{b} \in B$. Hence

$$\|\mathbf{x}\| = \|\lambda \mathbf{a} + \mu \mathbf{b}\| \leq |\lambda| \|\mathbf{a}\| + |\mu| \|\mathbf{b}\| \leq |\lambda| r_1 + |\mu| r_2.$$

This shows that $\lambda A + \mu B$ is bounded.

- (iv) This follows immediately from (ii) and (iii).

Subsection 9

Continuity

Continuity

- Let $f : A \rightarrow \mathbb{R}^m$ be a mapping, where A is a non-empty set in \mathbb{R}^n .
- Then f is said to be **continuous at a point \mathbf{a}** of A if, for each sequence $\mathbf{a}_1, \dots, \mathbf{a}_k, \dots$ of points of A that converges to \mathbf{a} , the sequence $f(\mathbf{a}_1), \dots, f(\mathbf{a}_k), \dots$ of points of \mathbb{R}^m converges to $f(\mathbf{a})$.
- If f is continuous at all points of A , then f is said to be **continuous on A** .
- An important example of a continuous mapping is the norm mapping $\|\cdot\| : \mathbb{R}^n \rightarrow \mathbb{R}$ defined by the equation $\|\cdot\|(\mathbf{x}) = \|\mathbf{x}\|$ for each point \mathbf{x} of \mathbb{R}^n .

That $\|\cdot\|$ is continuous follows immediately from the fact that $\|\mathbf{a}_k\| \rightarrow \|\mathbf{a}\|$ as $k \rightarrow \infty$ whenever $\mathbf{a}_k \rightarrow \mathbf{a}$ as $k \rightarrow \infty$.

Lipschitz Condition

- A mapping $f : A \rightarrow \mathbb{R}^m$ defined on a non-empty set A in \mathbb{R}^n is said to satisfy a **Lipschitz condition** on A if there exists a real number s such that, for all $\mathbf{x}, \mathbf{y} \in A$,

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

- If $f : A \rightarrow \mathbb{R}^m$ satisfies the Lipschitz condition, then it is continuous on A .

To see this, suppose that $\mathbf{a}_1, \dots, \mathbf{a}_k, \dots$ is a sequence of points of A that converges to a point \mathbf{a} of A , so that $\|\mathbf{a}_k - \mathbf{a}\| \rightarrow 0$ as $k \rightarrow \infty$.

The Lipschitz condition shows that

$$\|f(\mathbf{a}_k) - f(\mathbf{a})\| \leq s\|\mathbf{a}_k - \mathbf{a}\|.$$

Hence, $\|f(\mathbf{a}_k) - f(\mathbf{a})\| \rightarrow 0$ as $k \rightarrow \infty$, i.e., the sequence $f(\mathbf{a}_1), \dots, f(\mathbf{a}_k), \dots$ converges to $f(\mathbf{a})$.

Since f is continuous at an arbitrary \mathbf{a} of A , f is continuous on A .

Affine Transformations are Lipschitz Mappings

- The norm mapping $\|\cdot\| : \mathbb{R}^n \rightarrow \mathbb{R}$ considered above satisfies the Lipschitz condition.
- Every affine transformation $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$ satisfies a Lipschitz condition on \mathbb{R}^n .

Suppose that $\mathbf{Q} = [q_{ij}]$ is the real $m \times n$ matrix and \mathbf{q} the real $m \times 1$ matrix such that, for each vector \mathbf{x} in \mathbb{R}^n , considered as a column vector, $T(\mathbf{x}) = \mathbf{Q}\mathbf{x} + \mathbf{q}$. Let $\mathbf{x}, \mathbf{y} \in \mathbb{R}^n$. Write $\mathbf{u} = (u_1, \dots, u_n) = \mathbf{x} - \mathbf{y}$. By the Cauchy-Schwarz inequality, for $i = 1, \dots, m$,

$$(q_{i1}u_1 + \dots + q_{in}u_n)^2 \leq (q_{i1}^2 + \dots + q_{in}^2)(u_1^2 + \dots + u_n^2).$$

Setting $s = \sqrt{\sum_{i=1}^m \sum_{j=1}^n q_{ij}^2}$, we get

$$\begin{aligned} \|T(\mathbf{x}) - T(\mathbf{y})\|^2 &= \|\mathbf{Q}\mathbf{u}\|^2 = \sum_{i=1}^m (q_{i1}u_1 + \dots + q_{in}u_n)^2 \\ &\leq \sum_{i=1}^m (q_{i1}^2 + \dots + q_{in}^2)(u_1^2 + \dots + u_n^2) \\ &= s^2 \|\mathbf{u}\|^2. \end{aligned}$$

The Distance Function

- The **distance function** $d_A : \mathbb{R}^n \rightarrow \mathbb{R}$ of a non-empty set A in \mathbb{R}^n satisfies a Lipschitz condition.
- This function d_A associates with each point \mathbf{x} of \mathbb{R}^n its distance $d_A(\mathbf{x})$ from A .
- Formally, d_A is defined by the equation

$$d_A(\mathbf{x}) = \inf \{ \|\mathbf{x} - \mathbf{a}\| : \mathbf{a} \in A \}, \quad \text{for } \mathbf{x} \in \mathbb{R}^n.$$

- If A is the singleton set $\{\mathbf{a}\}$, then $d_A(\mathbf{x}) = \|\mathbf{x} - \mathbf{a}\|$.
- In particular, if $\mathbf{a} = \mathbf{0}$, then $d_A(\mathbf{x}) = \|\mathbf{x}\|$.
- It follows from the definition of d_A and a previous theorem that a point \mathbf{x} of \mathbb{R}^n lies in the closure $\text{cl}A$ of A if and only if its distance $d_A(\mathbf{x})$ from A is zero.

The Distance Function is Lipschitz

- Suppose now that \mathbf{x}, \mathbf{y} lie in \mathbb{R}^n .
- Then, for each $\varepsilon > 0$, there exists \mathbf{a} in A such that $\|\mathbf{x} - \mathbf{a}\| < d_A(\mathbf{x}) + \varepsilon$.
- By the triangle inequality,

$$d_A(\mathbf{y}) \leq \|\mathbf{y} - \mathbf{a}\| \leq \|\mathbf{y} - \mathbf{x}\| + \|\mathbf{x} - \mathbf{a}\| < \|\mathbf{y} - \mathbf{x}\| + d_A(\mathbf{x}) + \varepsilon.$$

- Since $\varepsilon > 0$ is arbitrary, $d_A(\mathbf{y}) \leq \|\mathbf{y} - \mathbf{x}\| + d_A(\mathbf{x})$.
- Interchanging \mathbf{x} and \mathbf{y} in this inequality, $d_A(\mathbf{x}) \leq \|\mathbf{x} - \mathbf{y}\| + d_A(\mathbf{y})$.
- Hence d_A satisfies the Lipschitz condition

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

- It follows that d_A is continuous on \mathbb{R}^n .

Remark

- In general, the \inf in the definition of d_A cannot be replaced by \min .
- To see this, suppose that A is the set $\mathbb{R}^n \setminus \{\mathbf{0}\}$.
Then $d_A(\mathbf{0}) = 0$, but there is no $\mathbf{a} \in A$ such that $\|\mathbf{0} - \mathbf{a}\| = 0$.

Distance from Nonempty Closed Sets

Theorem

Let A be a non-empty closed set in \mathbb{R}^n and let $\mathbf{x} \in \mathbb{R}^n$. Then there exists $\mathbf{a}_0 \in A$ such that $d_A(\mathbf{x}) = \|\mathbf{x} - \mathbf{a}_0\|$.

- It follows easily from the definition of $d_A(\mathbf{x})$ that there exists a sequence $\mathbf{a}_1, \dots, \mathbf{a}_k, \dots$ of points of A such that $\|\mathbf{x} - \mathbf{a}_k\| \rightarrow d_A(\mathbf{x})$ as $k \rightarrow \infty$. Since convergent sequences in \mathbb{R} are bounded, there exists a real number r such that $\|\mathbf{x} - \mathbf{a}_k\| \leq r$ for $k = 1, 2, \dots$. We have

$$\|\mathbf{a}_k\| \leq \|\mathbf{a}_k - \mathbf{x}\| + \|\mathbf{x}\| \leq r + \|\mathbf{x}\|, \text{ for } k = 1, 2, \dots$$

So the sequence $\mathbf{a}_1, \dots, \mathbf{a}_k, \dots$ is bounded. Hence it contains some subsequence $\mathbf{a}_{i_1}, \dots, \mathbf{a}_{i_k}, \dots$ which converges to a point \mathbf{a}_0 of \mathbb{R}^n . Since A is closed, $\mathbf{a}_0 \in A$. Now $\|\mathbf{x} - \mathbf{a}_{i_k}\| \rightarrow \|\mathbf{x} - \mathbf{a}_0\|$ as $k \rightarrow \infty$. But we already know that $\|\mathbf{x} - \mathbf{a}_{i_k}\| \rightarrow d_A(\mathbf{x})$ as $k \rightarrow \infty$. The uniqueness of limits in \mathbb{R} shows that $d_A(\mathbf{x}) = \|\mathbf{x} - \mathbf{a}_0\|$.

- The point \mathbf{a}_0 is called a **nearest point** of A to \mathbf{x} .

Continuity and Compactness

Theorem

Let $f : A \rightarrow \mathbb{R}^n$ be a continuous mapping, where A is a non-empty compact set in \mathbb{R}^n . Then $f(A)$ is a compact set in \mathbb{R}^n .

- Let $f(\mathbf{a}_1), \dots, f(\mathbf{a}_k), \dots$ be a sequence of points of $f(A)$, where $\mathbf{a}_1, \dots, \mathbf{a}_k, \dots$ is a sequence of points of A .

Since A is compact, there is a subsequence $\mathbf{a}_{i_1}, \dots, \mathbf{a}_{i_k}, \dots$ of $\mathbf{a}_1, \dots, \mathbf{a}_k, \dots$ which converges to some point \mathbf{a} of A .

By the continuity of f , the subsequence $f(\mathbf{a}_{i_1}), \dots, f(\mathbf{a}_{i_k}), \dots$ of $f(\mathbf{a}_1), \dots, f(\mathbf{a}_k), \dots$ converges to the point $f(\mathbf{a})$ of $f(A)$.

Thus $f(A)$ is compact.

Attainability of Sup and Inf

- Recall from elementary analysis that a continuous function $f : [a, b] \rightarrow \mathbb{R}$ is bounded and attains its bounds.

Corollary

Let $f : A \rightarrow \mathbb{R}$ be a continuous mapping, where A is a non-empty compact set in \mathbb{R}^n . Then there exist $\mathbf{a}, \mathbf{b} \in A$ such that

$$f(\mathbf{a}) = \inf \{f(\mathbf{x}) : \mathbf{x} \in A\} \quad \text{and} \quad f(\mathbf{b}) = \sup \{f(\mathbf{x}) : \mathbf{x} \in A\}.$$

- The theorem shows that the non-empty set $f(A) = \{f(\mathbf{x}) : \mathbf{x} \in A\}$ of real numbers is compact, and therefore closed and bounded. Thus $f(A)$ possesses both an infimum and supremum. Moreover, the infimum and supremum of $f(A)$ belong to $\text{cl} f(A)$. Hence, since $f(A)$ is closed, they belong to $f(A)$. So there exist $\mathbf{a}, \mathbf{b} \in A$ such that $f(\mathbf{a}) = \inf f(A)$ and $f(\mathbf{b}) = \sup f(A)$.

Attainability of Infimum of Distance

Theorem

Let A and B be non-empty sets in \mathbb{R}^n with A closed and B compact. Then there exist $\mathbf{a}_0 \in A$, $\mathbf{b}_0 \in B$ such that

$$\|\mathbf{a}_0 - \mathbf{b}_0\| = \inf \{\|\mathbf{a} - \mathbf{b}\| : \mathbf{a} \in A, \mathbf{b} \in B\}.$$

- The distance function d_A of A is continuous on \mathbb{R}^n . So, by restriction, it is continuous on B . By the corollary, applicable since B is compact, there exists $\mathbf{b}_0 \in B$ such that $d_A(\mathbf{b}_0) = \inf \{d_A(\mathbf{b}) : \mathbf{b} \in B\}$. By a previous theorem, applicable since A is closed, there exists $\mathbf{a}_0 \in A$ such that $d_A(\mathbf{b}_0) = \|\mathbf{b}_0 - \mathbf{a}_0\|$. For each $\mathbf{a} \in A$, $\mathbf{b} \in B$, we have

$$\|\mathbf{a} - \mathbf{b}\| \geq d_A(\mathbf{b}) \geq d_A(\mathbf{b}_0) = \|\mathbf{a}_0 - \mathbf{b}_0\|.$$

Since $\mathbf{a}_0 \in A$, $\mathbf{b}_0 \in B$, $\|\mathbf{a}_0 - \mathbf{b}_0\| = \inf \{\|\mathbf{a} - \mathbf{b}\| : \mathbf{a} \in A, \mathbf{b} \in B\}$.

- We refer to \mathbf{a}_0 and \mathbf{b}_0 as **nearest points** of A and B .

Continuity and Positivity

- Recall that if a real function is both continuous and positive at some point, then it is positive at all points of its domain sufficiently close to that point.

Theorem

Let the mapping $f : A \rightarrow \mathbb{R}$ be both continuous and positive at some point \mathbf{a} of a set A in \mathbb{R}^n . Then there exists an $r > 0$ such that $f(\mathbf{x}) > 0$ whenever $\mathbf{x} \in B(\mathbf{a}; r) \cap A$.

- Suppose that the stated conclusion does not hold. Then, for each $k = 1, 2, \dots$ there exists $\mathbf{a}_k \in B(\mathbf{a}; \frac{1}{k}) \cap A$ such that $f(\mathbf{a}_k) \leq 0$. Since f is continuous at \mathbf{a} and $\mathbf{a}_k \rightarrow \mathbf{a}$ as $k \rightarrow \infty$, $f(\mathbf{a}_k) \rightarrow f(\mathbf{a})$ as $k \rightarrow \infty$. Because $f(\mathbf{a}_k) \leq 0$ for $k = 1, 2, \dots$, it follows that $f(\mathbf{a}) \leq 0$. This contradiction establishes the theorem.

Continuity of Composition

- Recall that a continuous function of a continuous function is itself continuous.

Theorem

Let $f : A \rightarrow \mathbb{R}^m$ and $g : B \rightarrow \mathbb{R}^p$ be continuous mappings, where A and B are, respectively, non-empty sets in \mathbb{R}^n and \mathbb{R}^m such that $f(A) \subseteq B$. Then the composite mapping $g \circ f : A \rightarrow \mathbb{R}^p$ is continuous.

- Let $\mathbf{a}_1, \dots, \mathbf{a}_k, \dots$ be a sequence of points of A that converges to a point \mathbf{a} of A . Since f is continuous, the sequence of points $f(\mathbf{a}_1), \dots, f(\mathbf{a}_k), \dots$ of B converges to the point $f(\mathbf{a})$ of B . Since g is continuous, the sequence $g(f(\mathbf{a}_1)), \dots, g(f(\mathbf{a}_k)), \dots$ converges to $g(f(\mathbf{a}))$, i.e., the sequence $(g \circ f)(\mathbf{a}_1), \dots, (g \circ f)(\mathbf{a}_k), \dots$ converges to $(g \circ f)(\mathbf{a})$. This shows that $g \circ f$ is continuous.

Inverse Images of Open and of Closed Sets

Theorem

Let $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$ be a continuous mapping and let B be a closed (open) subset of \mathbb{R}^m . Then $f^{-1}(B)$ is closed (open).

- Suppose first that B is closed. Let $\mathbf{a}_1, \dots, \mathbf{a}_k, \dots$ be a sequence of points of $f^{-1}(B)$ that converges to a point \mathbf{a} of \mathbb{R}^n . The continuity of f shows that the sequence of points $f(\mathbf{a}_1), \dots, f(\mathbf{a}_k), \dots$ of B converges to the point $f(\mathbf{a})$ of \mathbb{R}^m . But B is closed. So $f(\mathbf{a}) \in B$, i.e., $\mathbf{a} \in f^{-1}(B)$. This shows that $f^{-1}(B)$ is closed.

Suppose next that B is open. Then the complement B^c of B in \mathbb{R}^m is closed. Hence, by what has just been proved, $f^{-1}(B^c)$ is closed in \mathbb{R}^n . Thus, the complement $f^{-1}(B) = (f^{-1}(B^c))^c$ in \mathbb{R}^n is open.