

1.3 Euclidean Space

Product spaces are ubiquitous in Mathematics and its applications.

Def: The Cartesian Product of sets A & B is $A \times B$, defined as:

$$A \times B = \{(a, b) \mid a \in A, b \in B\}$$

We will write $\mathbb{R} \times \mathbb{R}$ as \mathbb{R}^2 , $\mathbb{R} \times \mathbb{R} \times \mathbb{R}$ as \mathbb{R}^3 , and so on

Note that \mathbb{R}^m is a vector field over \mathbb{R} — this means that addition is distributive, associative, and invertible, and scalar multiplication likewise.

Def: The Inner Product, or Dot Product of two vectors $\mathbf{x}, \mathbf{y} \in \mathbb{R}^m$ is:

$$\langle \mathbf{x}, \mathbf{y} \rangle = x_1 y_1 + x_2 y_2 + \dots + x_m y_m = \mathbf{x} \cdot \mathbf{y}.$$

Thm: The inner product $\langle \cdot, \cdot \rangle$ on \mathbb{R}^m satisfies:

$$\langle \mathbf{x}, \mathbf{y} + c\mathbf{z} \rangle = \langle \mathbf{x}, \mathbf{y} \rangle + c\langle \mathbf{x}, \mathbf{z} \rangle, \quad [\text{Bilinearity}]$$

$$\langle \mathbf{x}, \mathbf{y} \rangle = \langle \mathbf{y}, \mathbf{x} \rangle, \quad [\text{Symmetry}] \text{ and}$$

$$\langle \mathbf{x}, \mathbf{x} \rangle \geq 0, \text{ with } \langle \mathbf{x}, \mathbf{x} \rangle = 0 \text{ iff } \mathbf{x} = \mathbf{0}. \quad [\text{non-negative}]$$

Proof: L.T.S. \square

Def: The magnitude of \mathbf{x} is $|\mathbf{x}| = \sqrt{\langle \mathbf{x}, \mathbf{x} \rangle}$.

(Note that this corresponds with the magnitude of a real $\#$, and provides an alternate definition:

$$\text{For } x \in \mathbb{R}, \quad |x| = \sqrt{x^2}.)$$

Thm (Cauchy-Schwartz): For any $x, y \in \mathbb{R}^n$, $\langle x, y \rangle \leq |x| \cdot |y|$.

Proof: Define the function $Q: \mathbb{R} \rightarrow \mathbb{R}$ by $Q(t) = \langle x+ty, x+ty \rangle$.

Then by bilinearity, we have:

$$\begin{aligned} Q(t) &= \langle x+ty, x+ty \rangle = \langle x+ty, x \rangle + t \langle x+ty, y \rangle \\ &= \langle x, x \rangle + t \langle y, x \rangle + t (\langle x, y \rangle + t \langle y, y \rangle) \\ &= \langle y, y \rangle t^2 + 2t \langle x, y \rangle + \langle x, x \rangle \\ &= at^2 + bt + c \end{aligned}$$

Then the graph of $Q(t)$ is a parabola.

Note that since $Q(t)$ is the value of an inner product, we have $Q(t) \geq 0 \forall t$. In particular, there can be at most one real root of the equation $Q(t) = 0$.

This means the discriminant must be nonpositive, i.e.:

$$b^2 - 4ac = (2\langle x, y \rangle)^2 - 4\langle y, y \rangle \langle x, x \rangle \leq 0.$$

Rearranging, and cancelling a 4, gives:

$$\langle x, y \rangle^2 \leq \langle x, x \rangle \langle y, y \rangle, \text{ or}$$

$$\langle x, y \rangle \leq \sqrt{\langle x, x \rangle} \sqrt{\langle y, y \rangle} = |x| |y|. \quad \square$$

Pedantic Note: What if $y = 0$?

Cor: The triangle inequality: $|x+y| \leq |x|+|y| \quad \forall x, y \in \mathbb{R}^m$.

Proof: Consider instead $|x+y|^2$:

$$|x+y|^2 = \langle x+y, x+y \rangle = \langle x, x \rangle + 2\langle x, y \rangle + \langle y, y \rangle$$

$$\leq |x|^2 + 2|x||y| + |y|^2 = (|x|+|y|)^2$$

Taking the square root of both sides gives:

$$|x+y| \leq |x|+|y|, \text{ as required. } \blacksquare$$

Def: The Euclidean Distance between x and y in \mathbb{R}^m is:

$$|x-y| = \sqrt{\langle x-y, x-y \rangle} = \sqrt{(x_1-y_1)^2 + \dots + (x_m-y_m)^2}$$

Note: We will soon be considering other distances, so don't drop the 'Euclidean'!

Prop: For all $x, y, z \in \mathbb{R}^m$, $|x-z| \leq |x-y| + |y-z|$.

(Actual triangles are involved this time!)

Proof: LTS. \square

Some common sets in \mathbb{R}^m :

- For each $i=1, \dots, m$, let $I_i = [a_i, b_i] \subseteq \mathbb{R}$. Then:

$I_1 \times I_2 \times \dots \times I_m$ is a box in \mathbb{R}^m .

- The unit cube is $[0, 1]^m$.

- The unit ball in \mathbb{R}^m is:

$$B^m = \{x \in \mathbb{R}^m \mid |x| \leq 1\}$$

- The unit sphere is the boundary of B^m , $S^{m-1} = \{x \in \mathbb{R}^m \mid |x| = 1\}$.

Aside: What do you think happens when we consider \mathbb{R}^∞ ?
The answer may surprise you!

In optimization problems, it is often important that the set of potential solutions (the feasible set) is convex:

Def. $E \subseteq \mathbb{R}^m$ is convex if for all $x, y \in \mathbb{R}^m$, the line segment \overline{xy} is contained in E . In other words, if:

$$(1-t)x + ty \in E \text{ for all } t \in [0, 1].$$

This is called a convex combination of x & y .

Prop. Every box in \mathbb{R}^m is convex.

B^m is convex.

S^{m-1} is not convex.

Proof. We show the second; the first and third are L.T.S.

Let $x, y \in B^m$. Then $|x| \leq 1$ and $|y| \leq 1$.

Let $z = (1-t)x + ty$. We have:

$$\begin{aligned}
 |z|^2 &= \langle z, z \rangle = (1-t)^2 \langle x, x \rangle + 2(1-t)t \langle x, y \rangle + t^2 \langle y, y \rangle \\
 &\leq (1-t)^2 |x|^2 + 2(1-t)t |x| |y| + t^2 |y|^2 \\
 &\leq (1-t)^2 + 2(1-t)t + t^2 = (1-t+t)^2 = 1
 \end{aligned}$$

But $|z|^2 \leq 1 \implies |z| \leq 1. \quad \square$

Other inner product spaces exist; consider: $C([a, b])$, the space of continuous, real-valued functions on $[a, b]$.

For $f, g \in C([a, b])$, define: $\langle f, g \rangle = \int_a^b f(x)g(x)dx$.

Given an inner product $\langle \cdot, \cdot \rangle$, one can always define

a norm on the space by: $\|x\| = \sqrt{\langle x, x \rangle}$.

(A norm is a real-valued function satisfying:

- $\|x\| \geq 0$; $\|x\| = 0 \implies x = 0$
- $\|ax\| = |a| \|x\|$
- $\|x+y\| \leq \|x\| + \|y\|$

)

But not all norms come from inner products!

We will explore this idea again much later.