

Session 4: Norms and inner-products

Optimization and Computational Linear Algebra for Data Science

Contents

1. Why do we care about all these things ?
Application to data science: image compression
2. Norms & inner-products
3. Orthogonality
4. Orthogonal projection
5. Proof of the Cauchy-Schwarz inequality

Orthogonality

Definition

Definition (Orthogonality)

- ❖ We say that vectors x and y are *orthogonal* if $\langle x, y \rangle = 0$. We write then $x \perp y$.
- ❖ We say that a vector x is orthogonal to a set of vectors $A \subset V$ if x is orthogonal to all the vectors in A , i.e. $\forall y \in A, \langle x, y \rangle = 0$. We write then $x \perp A$.
- ❖ More generality we say that $A \subset V$ and $B \subset V$ are orthogonal if $\langle x, y \rangle = 0$ for all $x \in A$ and all $y \in B$. As before, we write $A \perp B$.

If x is orthogonal to v_1, \dots, v_k then x is orthogonal to any linear combination of these vectors i.e. $x \perp \text{Span}(v_1, \dots, v_k)$.

Pythagorean Theorem

Theorem (Pythagorean theorem)

Let $x, y \in V$. Then

$$x \perp y \iff \|x + y\|^2 = \|x\|^2 + \|y\|^2.$$

Proof.



Orthogonal & orthonormal families

Definition

Let v_1, \dots, v_k be vectors of V . We say that the family of vectors (v_1, \dots, v_k) is

- *orthogonal* if the vectors v_1, \dots, v_n are pairwise orthogonal, i.e. $\langle v_i, v_j \rangle = 0$ for all $i \neq j$.
- *orthonormal* if it is orthogonal and if all the v_i have unit norm: $\|v_1\| = \dots = \|v_k\| = 1$.

A toy example

Orthonormal basis are particularly convenient for computing coordinates of vectors:

Proposition

Assume that $\dim(V) = n$ and let (v_1, \dots, v_n) be an **orthonormal** basis of V . Then the coordinates of a vector $x \in V$ in the basis (v_1, \dots, v_n) are $(\langle v_1, x \rangle, \dots, \langle v_n, x \rangle)$:

$$x = \langle v_1, x \rangle v_1 + \cdots + \langle v_n, x \rangle v_n.$$

Moreover, for all $y \in V$, we have

$\langle x, y \rangle = \langle v_1, x \rangle \langle v_1, y \rangle + \cdots + \langle v_n, x \rangle \langle v_n, y \rangle$. Taking $y = x$ leads to

$$\|x\| = \sqrt{\langle v_1, x \rangle^2 + \cdots + \langle v_n, x \rangle^2}.$$

Questions?