

# Lecture 18: Inner Product Spaces

March 19, 2024

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# Inner Products

In  $\mathbb{R}^2$ , the scalar (or dot) product of the vectors  $x = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$  and  $y = \begin{pmatrix} y_1 \\ y_2 \end{pmatrix}$  is given by

$$x \cdot y = x_1 y_1 + x_2 y_2 = x^T y = y^T x = y \cdot x.$$

We can interpret the *length*  $\|x\|$  of the vector  $x$  as the length of the directed line segment from the origin to  $(x_1, x_2)$ , which by the Theorem of Pythagoras is  $\sqrt{x_1^2 + x_2^2}$  or  $\sqrt{x \cdot x}$ .

Once we have a concept of **length** of a vector, we can define the *distance*  $d(x, y)$  between two vectors  $x$  and  $y$  as the length of their difference:  
 $d(x, y) = \|x - y\|.$

# Inner Products

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Similarly, from the Cosine Rule we can observe that

$x \cdot y = \|x\| \|y\| \cos \theta$ , where  $\theta$  is the angle between the directed line segments representing  $x$  and  $y$ . In particular,  $x$  is orthogonal to  $y$  (or  $x \perp y$ ) if and only if  $x \cdot y = 0$ .

So the scalar product encodes geometric information in  $\mathbb{R}^2$ , and it also provides a mechanism for defining concepts of length, distance and orthogonality on real vector spaces that do not necessarily have an obvious geometric structure.

The scalar product is an example of an [inner product](#).

An **inner product** on a vector space  $V$  is a function from  $V \times V$  to  $\mathbb{R}$  that assigns an element of  $\mathbb{R}$  to every ordered pair of elements of  $V$ , and has the following properties.

- 1 Symmetry:  $\langle x, y \rangle = \langle y, x \rangle$  for all  $x, y \in V$
- 2 Linearity in both slots (bilinearity): For all  $x, y, z \in V$  and all  $a, b \in \mathbb{R}$ , we have  $\langle ax + by, z \rangle = a\langle x, z \rangle + b\langle y, z \rangle$  and  $\langle x, ay + bz \rangle = a\langle x, y \rangle + b\langle x, z \rangle$ .
- 3 Non-negativity:  $\langle x, x \rangle \geq 0$  for all  $x \in V$ , and  $\langle x, x \rangle = 0$  only if  $x = 0_V$ .

The ordinary scalar product on  $\mathbb{R}^n$  is the best known example of an inner product.

# Examples of inner products

- 1 The ordinary scalar product on  $\mathbb{R}^n$ .
- 2 Let  $C$  be the vector space of all continuous real-valued functions on the interval  $[0, 1]$ . The analogue of the ordinary scalar product on  $C$  is the inner product given by

$$\langle f, g \rangle = \int_0^1 f(x)g(x) dx, \text{ for } f, g \in C.$$

- 3 On the space  $M_{m \times n}(\mathbb{R})$ , the *Frobenius inner product* or *trace inner product* is defined by  $\langle A, B \rangle = \text{trace}(A^T B)$ . Note that  $\text{trace} A^T B$  is the sum over all positions  $(i, j)$  of the products  $A_{ij}B_{ij}$ . So this is closely related to the ordinary scalar product, if the matrices  $A$  and  $B$  were regarded as vectors with  $mn$  entries over  $\mathbb{R}$ .

It is possible for a single vector space to have many different inner products defined on it, and if there is any risk of ambiguity we need to specify which one we are considering.

# Length, Distance and Orthogonality

Given a real vector space and equipped with an inner product  $\langle \cdot, \cdot \rangle$ , we make the following two definitions.

**Definition** We define the **length** or **norm** of any vector  $v$  by

$$\|v\| = \sqrt{\langle v, v \rangle},$$

and we define the **distance** between the vectors  $u$  and  $v$  by

$$d(u, v) = \|u - v\|.$$

**Definition** We say that vectors  $u$  and  $v$  are **orthogonal** (with respect to  $\langle \cdot, \cdot \rangle$ ) if  $\langle u, v \rangle = 0$ .

These definitions are consistent with “typical” geometrically motivated concepts of distance and orthogonality.

# Unit Vectors and Scaling

An element  $v$  of  $V$  is referred to as a **unit vector** if  $\|v\| = 1$ .

The norm of elements of  $V$  has the property that  $\|cv\| = |c| \|v\|$  for any vector  $v$  and real scalar  $c$ . To see this we can note that

$$\|cv\| = \sqrt{\langle cv, cv \rangle} = \sqrt{c^2 \langle v, v \rangle} = |c| \|v\|.$$

So we can adjust the norm of any element of  $V$ , while preserving its direction, by multiplying it by a positive scalar.

**Definition** If  $v$  is a non-zero vector in an inner product space  $V$ , then

$$\hat{v} := \frac{1}{\|v\|} v$$

is a unit vector in the same direction as  $v$ , referred to as the *normalization* of  $v$ .