# Lecture 12: Inner Product Spaces

### Elif Tan

Ankara University

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### Definition (Inner Product Space)

Let  $(V, \oplus, \odot)$  be a real vector space. If the function  $\langle, \rangle : V \times V \to \mathbb{R}$ satisfies the following properties, then V is called an inner product space and the function  $\langle, \rangle$  is called an inner product function. (i)  $\forall u \in V, \langle u, u \rangle \ge 0$  and  $\langle u, u \rangle = 0 \Leftrightarrow u = 0$ (ii)  $\forall u, v \in V, \langle u, v \rangle = \langle v, u \rangle$ (iii)  $\forall u, v, w \in V, \langle u \oplus v, w \rangle = \langle u, w \rangle + \langle v, w \rangle$ (iv)  $\forall u, v \in V, \forall c \in \mathbb{R}, \langle c \odot u, v \rangle = c \langle u, v \rangle$ .

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## Example

For 
$$u = \begin{bmatrix} u_1 \\ u_2 \\ \vdots \\ u_n \end{bmatrix}$$
,  $v = \begin{bmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{bmatrix} \in \mathbb{R}^n$ , the standard inner product (dot product) on  $\mathbb{R}^n$  is defined by
$$\langle u, v \rangle = u_1 v_1 + u_2 v_2 + \dots + u_n v_n.$$

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#### Theorem

Let V be an inner product space, and  $S = \{u_1, u_2, ..., u_n\}$  be an ordered basis for the vector space V. Then the matrix  $A = [a_{ij}]_{n \times n}$ ,  $a_{ij} := \langle a_i, a_j \rangle$  is a symmetric matrix, and for every  $u, v \in V$ , it determines  $\langle u, v \rangle$ .

Note that the matrix  $A = [a_{ij}]_{n \times n}$ ,  $a_{ij} = \langle u_i, u_j \rangle$  is called the matrix of the inner product with respect to the ordered basis *S*.

$$A = \begin{bmatrix} \langle u_1, u_1 \rangle & \langle u_1, u_2 \rangle & \cdots & \langle u_1, u_n \rangle \\ \langle u_2, u_1 \rangle & \langle u_2, u_2 \rangle & \cdots & \langle u_2, u_n \rangle \\ \vdots & \vdots & \ddots & \vdots \\ \langle u_n, u_1 \rangle & \langle u_n, u_2 \rangle & \cdots & \langle u_n, u_n \rangle \end{bmatrix}$$

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## Definition (Positive definite matrix)

The  $n \times n$  symmetric matrix A is called positive definite matrix if it has the property that

$$\forall 0 \neq x \in \mathbb{R}^n$$
,  $x^T A x > 0$ .

### Theorem

Let  $A = [a_{ij}]_{n \times n}$  be a positive definite matrix, and  $S = \{u_1, u_2, \ldots, u_n\}$  be an ordered basis for the vector space V. Then the function  $\langle, \rangle : V \times V \to \mathbb{R}$  that is defined by  $\forall u, v \in V, \langle u, v \rangle := [u]_S^T A[v]_S$ , is an inner product function on V.

Note that it is not easy to determine when a symmetric matrix is positive definite!

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# Inner Product Spaces

### Example

Consider the standard inner product function  $\left\langle \begin{vmatrix} a \\ b \end{vmatrix}, \begin{vmatrix} c \\ d \end{vmatrix} \right\rangle = ac + bd$ on  $\mathbb{R}^2$ . The matrix of the inner product with respect to the ordered basis  $S = \left\{ u_1 = \left[ \begin{array}{c} 1 \\ 0 \end{array} \right], u_2 = \left[ \begin{array}{c} 0 \\ 1 \end{array} \right] \right\}$  is

$$A = \begin{bmatrix} \langle u_1, u_1 \rangle & \langle u_1, u_2 \rangle \\ \langle u_2, u_1 \rangle & \langle u_2, u_2 \rangle \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

Conversely, consider the matrix  $A = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$ . (A is positive definite matrix, verify it). The inner product with respect to the ordered standard basis in  $\mathbb{R}^2$  is

$$\left\langle \begin{bmatrix} a \\ b \end{bmatrix}, \begin{bmatrix} c \\ d \end{bmatrix} \right\rangle = \begin{bmatrix} a & b \end{bmatrix} A \begin{bmatrix} c \\ d \end{bmatrix} = ac + bd.$$

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## Definition (Lenght)

Let V be an inner product space. The lenght of  $v \in V$  is defined by

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

## Definition (Distance)

Let V be an inner product space. The distance between u and v in V is defined by

$$d(u,v) := ||u-v|| = \sqrt{\langle u-v, u-v \rangle}.$$

Theorem (Cauchy-Schwarz Inequality)

Let V be an inner product space.  $\forall u, v \in V$ ,

 $|\langle \mathbf{v}, \mathbf{v} \rangle| \leq ||\mathbf{u}|| \, ||\mathbf{v}|| \, .$ 

By using Cauchy-Schwarz inequality, we define the cosine of an angle between nonzero vectors u and v in V as

$$\cos heta:=rac{\langle m{v},m{v}
angle}{||m{u}||\,||m{v}||}, 0\leq heta\leq\pi.$$

## Corollary

Let V be an inner product space. Then  $\forall u, v \in V$  and  $\forall c \in \mathbb{R}$ ,

$$||c \odot v|| \le |c| ||v|| d(u, v) = 0 \Leftrightarrow u = v d(u, v) = d(v, u) ||u \oplus v|| \le ||u|| + ||v|| (Triangle inequality)$$

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### Definition

Let V be an inner product space. The vectors u and v in V are orthogonal if  $\langle v, v \rangle = 0$ . That is,

$$u \perp v \Leftrightarrow \langle v, v \rangle = 0.$$

A set of S of vectors in V is called ortogonal if any two distinct vectors in S are orthogonal. Additionally, if each vector in S is a unit vector  $(||u|| = 1, u \in V)$ , then S is called orthonormal.

#### Theorem

- Zero vector is orthogonal with every vector in an inner product space V, that is, 0 ⊥ v, ∀v ∈ V.
- If S = {u<sub>1</sub>, u<sub>2</sub>, ..., u<sub>n</sub>} is an orthogonal set of nonzero vectors in an inner product space V, then S is linearly independent.