

# **MATH 2135 Linear Algebra**

## Chapter 6 Inner Product Spaces

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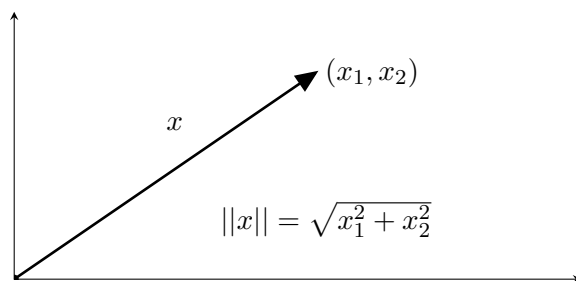
## Contents

<b>1</b>	<b>6.A Inner Products and Norms</b>	<b>3</b>
1.1	Definition of dot product . . . . .	3
1.2	Definition of inner product . . . . .	4
1.2.1	Examples . . . . .	5
1.3	Definition of inner product space . . . . .	5
1.4	Basic properties of an inner product . . . . .	6
1.5	Definition of norm, $\ v\ $ . . . . .	6
1.6	Basic properties of the norm . . . . .	7
1.7	Definition of orthogonal . . . . .	7
1.8	Orthogonality and 0 . . . . .	7
1.9	Pythagorean Theorem . . . . .	8
1.10	Orthogonal Decomposition (Projection) . . . . .	9
1.11	Cauchy-Schwarz Inequality . . . . .	10
1.11.1	Examples of the Cauchy-Schwarz Inequality . . . . .	11
1.12	Triangle Inequality . . . . .	11
1.13	Parallelogram Equality . . . . .	12
<b>2</b>	<b>6.B Orthonormal Bases</b>	<b>12</b>
<b>3</b>	<b>6.C Orthogonal Complements and Minimization Problems</b>	<b>12</b>

# 1 6.A Inner Products and Norms

To motivate the concept of inner product, think of vectors in  $\mathbb{R}^2$  and  $\mathbb{R}^3$  as arrows with initial point at the origin. The length of a vector  $x$  in  $\mathbb{R}^2$  or  $\mathbb{R}^3$  is called the *norm* of  $x$ , denoted  $\|x\|$ . Thus for  $x = (x_1, x_2) \in \mathbb{R}^2$ , we have  $\|x\| = \sqrt{x_1^2 + x_2^2}$ . The generalization to  $\mathbb{R}^n$  is: we defined the norm of  $x = (x_1, \dots, x_n) \in \mathbb{R}^n$  by

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



The norm is not linear on  $\mathbb{R}^n$ .

## 1.1 Definition of dot product

For  $x, y \in \mathbb{R}^n$ , the **dot product** of  $x$  and  $y$ , denoted  $x \cdot y$ , is defined by

$$x \cdot y = x_1 y_1 + \dots + x_n y_n$$

where  $x = (x_1, \dots, x_n)$  and  $y = (y_1, \dots, y_n)$ . Note that the dot product of two vectors in  $\mathbb{R}^n$  is a number, not a vector.

An inner product is a generalization of the dot product. Recall that if  $\lambda = a + bi$ , where  $a, b \in \mathbb{R}$ , then

- the absolute value of  $\lambda$ , denoted  $|\lambda|$ , is defined by  $|\lambda| = \sqrt{a^2 + b^2}$ ;
- the complex conjugate of  $\lambda$ , denoted  $\bar{\lambda}$ , is defined by  $\bar{\lambda} = a - bi$ ;
- $|\lambda|^2 = \lambda \bar{\lambda}$ .

For  $z = (z_1, \dots, z_n) \in \mathbb{C}^n$ , we define the norm of  $z$  by

$$\|z\| = \sqrt{|z_1|^2 + \dots + |z_n|^2}.$$

The absolute values are needed because we want  $\|z\|$  to be a nonnegative number. Note that

$$\|z\|^2 = z_1 \overline{z_1} + \cdots + z_n \overline{z_n}.$$

We want to think of  $\|z\|^2$  as the inner product of  $z$  with itself. The equation above suggests that the inner product of  $w = (w_1, \dots, w_n) \in \mathbb{C}^n$  with  $z$  should equal

$$w_1 \overline{z_1} + \cdots + w_n \overline{z_n}.$$

If the roles of  $w$  and  $z$  were interchanged, the expression above would be its complex conjugate. We should expect that the inner product of  $w$  with  $z$  equals the complex conjugate of the inner product of  $z$  with  $w$ .

## 1.2 Definition of inner product

An *inner product* on  $V$  is a function that takes each ordered pair  $(u, v)$  of elements of  $V$  to a number  $\langle u, v \rangle \in \mathbf{F}$  and has the following properties:

**positivity**

$$\langle v, v \rangle \geq 0 \text{ for all } v \in V;$$

**definiteness**

$$\langle v, v \rangle = 0 \text{ if and only if } v = 0;$$

**additivity in first slot**

$$\langle u + v, w \rangle = \langle u, w \rangle + \langle v, w \rangle \text{ for all } u, v, w \in V;$$

**homogeneity in first slot**

$$\langle \lambda u, v \rangle = \lambda \langle u, v \rangle \text{ for all } \lambda \in \mathbf{F} \text{ and all } u, v \in V;$$

**conjugate symmetry**

$$\langle u, v \rangle = \overline{\langle v, u \rangle} \text{ for all } u, v \in V.$$

Every real number equals its complex conjugate. If we are dealing with a real vector space, then the last condition can be  $\langle u, v \rangle = \langle v, u \rangle$  for all  $u, v \in V$ .

### 1.2.1 Examples

- (a) The **Euclidean inner product** on  $\mathbf{F}^n$  is defined by

$$\langle (w_1, \dots, w_n), (z_1, \dots, z_n) \rangle = w_1 \overline{z_1} + \dots + w_n \overline{z_n}.$$

- (b) If  $c_1, \dots, c_n$  are positive numbers, then an inner product can be defined on  $\mathbf{F}^n$  by

$$\langle (w_1, \dots, w_n), (z_1, \dots, z_n) \rangle = c_1 w_1 \overline{z_1} + \dots + c_n w_n \overline{z_n}.$$

- (c) An inner product can be defined on the vector space of continuous real-valued functions on the interval  $[-1, 1]$  by

$$\langle f, g \rangle = \int_{-1}^1 f(x) \overline{g(x)} dx.$$

This is an inner product since for example: additivity in the left slot is defined as

$$\begin{aligned} \langle f + h, g \rangle &= \int_{-1}^1 (f(x) + h(x)) \overline{g(x)} dx \\ &= \int_{-1}^1 f(x) \overline{g(x)} dx + \int_{-1}^1 h(x) \overline{g(x)} dx \\ &= \langle f, g \rangle + \langle h, g \rangle. \end{aligned}$$

- (d) An inner product can be defined on  $\mathcal{P}(\mathbb{R})$  by

$$\langle p, q \rangle = \int_0^\infty p(x) q(x) e^{-x} dx.$$

- (e) The dot product on  $\mathbb{R}^n$

$$\langle v, w \rangle = v \cdot w = x_1 y_1 + \dots + x_n y_n$$

and

$$\langle v, v \rangle = v \cdot v = x_1^2 + \dots + x_n^2 \geq 0.$$

### 1.3 Definition of inner product space

An **inner product space** is a vector space  $V$  along with an inner product on  $V$ . For the rest of this chapter,  $V$  denotes an inner product space over  $\mathbf{F}$ .

## 1.4 Basic properties of an inner product

- (a) For each fixed  $u \in V$ , the function that takes  $v$  to  $\langle v, u \rangle$  is a linear map from  $V$  to  $\mathbf{F}$ .

*Proof.*     •  $f(v + v') = \langle v + v', u \rangle = \langle v, u \rangle + \langle v', u \rangle = f(v) + f(v')$   
              •  $f(\lambda v) = \dots = \lambda f(v)$ .

□

- (b)  $\langle 0, u \rangle = 0$  for every  $u \in V$ .  
 (c)  $\langle u, 0 \rangle = 0$  for every  $u \in V$ .  
 (d)  $\langle u, v + w \rangle = \langle u, v \rangle + \langle u, w \rangle$  for all  $u, v, w \in V$ .

*Proof.* This is additivity in the second slot.

$$\begin{aligned} \langle u, v + w \rangle &= \overline{\langle v + w, u \rangle} \\ &= \overline{\langle v, u \rangle + \langle w, u \rangle} \\ &= \overline{\langle v, u \rangle} + \overline{\langle w, u \rangle} \\ &= \langle u, v \rangle + \langle u, w \rangle. \end{aligned}$$

□

- (e)  $\langle u, \lambda v \rangle = \bar{\lambda} \langle u, v \rangle$  for all  $\lambda \in \mathbf{F}$  and  $u, v \in V$ .

*Proof.* This is homogeneity in the second slot.

$$\begin{aligned} \langle u, \lambda v \rangle &= \overline{\langle \lambda v, u \rangle} \\ &= \overline{\lambda \langle v, u \rangle} \\ &= \bar{\lambda} \overline{\langle v, u \rangle} \\ &= \bar{\lambda} \langle u, v \rangle. \end{aligned}$$

□

## 1.5 Definition of norm, $\|v\|$

For  $v \in V$ , the **norm** of  $v$ , denoted  $\|v\|$ , is defined by

$$\|v\| = \sqrt{\langle v, v \rangle} \geq 0.$$

Note that  $\|v\|^2 = \langle v, v \rangle$ .

## 1.6 Basic properties of the norm

Suppose  $v \in V$ .

- (a)  $\|v\| = 0$  if and only if  $v = 0$ .
- (b)  $\|\lambda v\| = |\lambda|\|v\|$  for all  $\lambda \in \mathbf{F}$ .

*Proof.* (a) The desired result holds because  $\langle v, v \rangle = 0$  if and only if  $v = 0$ .

- (b) Suppose  $\lambda \in \mathbf{F}$ . then

$$\begin{aligned}\|\lambda v\|^2 &= \langle \lambda v, \lambda v \rangle \\ &= \lambda \langle v, \lambda v \rangle \\ &= \lambda \bar{\lambda} \langle v, v \rangle \\ &= |\lambda|^2 \|v\|^2.\end{aligned}$$

Taking square roots now gives the desired equality.

□

## 1.7 Definition of orthogonal

Two vectors  $u, v \in V$  are called **orthogonal** if  $\langle u, v \rangle = 0$ . We write  $u \perp v$  to mean “ $u$  is orthogonal to  $v$ .”

## 1.8 Orthogonality and 0

- (a) 0 is orthogonal to every vector in  $V$ .
- (b) 0 is the only vector in  $V$  that is orthogonal to itself.

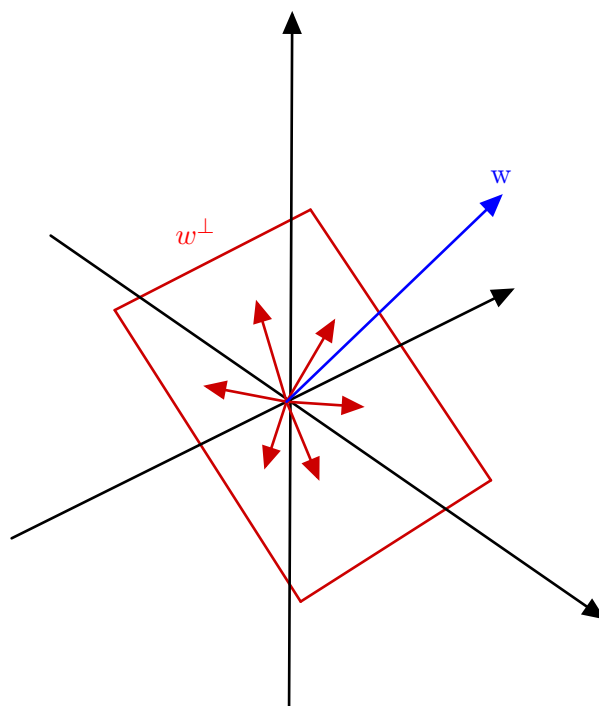
*Proof.* If  $v \in V$  and  $\langle v, v \rangle = 0$ , then  $v = 0$  (by definition of inner product). □

- (c)  $u \perp v \Leftrightarrow v \perp u$
- (d)  $u \perp w$  and  $v \perp w \Rightarrow (u + v) \perp w$ .
- (e)  $u \perp w$  and  $\lambda \in \mathbf{F} \Rightarrow (\lambda u) \perp w$ .

The last two properties imply that the set

$$w^\perp = \{v \mid v \perp w\}$$

is a subspace of  $V$ , called the *orthogonal complement* of  $V$ .



## 1.9 Pythagorean Theorem

Suppose  $u$  and  $v$  are orthogonal vectors in  $V$ . Then

$$\|u + v\|^2 = \|u\|^2 + \|v\|^2.$$



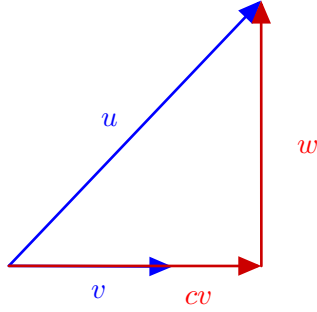
*Proof.* We have

$$\begin{aligned}
||u + v||^2 &= \langle u + v, u + v \rangle \\
&= \langle u, u + v \rangle + \langle v, u + v \rangle \\
&= \langle u, u \rangle + \underbrace{\langle u, v \rangle + \langle v, u \rangle}_0 + \langle v, v \rangle \\
&= \langle u, u \rangle + \langle v, v \rangle \\
&= ||u||^2 + ||v||^2,
\end{aligned}$$

as desired.  $\square$

### 1.10 Orthogonal Decomposition (Projection)

Given  $u, v \in V$ , assuming  $v \neq 0$ . Then we can write  $u$  as a sum of two vectors, the first of which is parallel to  $v$  and the second is orthogonal to  $v$ .



Let  $c = \frac{\langle u, v \rangle}{||v||^2} = \frac{\langle u, v \rangle}{\langle v, v \rangle}$  and let  $w = u - cv$ . Then  $\langle w, v \rangle = 0$  and  $u = cv + w$ .

*Proof.* We know  $u = cv + w$  holds by the definition of  $w$ . We also know that  $cv$  is parallel to  $v$  by the definition of “parallel.” To prove that  $w$  is orthogonal to  $v$ , we can calculate:

$$\begin{aligned}
\langle w, v \rangle &= \langle u - cv, v \rangle \\
&= \langle u, v \rangle - c\langle v, v \rangle \\
&= \langle u, v \rangle - \frac{\langle u, v \rangle}{\langle v, v \rangle} \langle v, v \rangle \\
&= \langle u, v \rangle - \langle u, v \rangle = 0.
\end{aligned}$$

Therefore,  $w \perp v$ . □

### 1.11 Cauchy-Schwarz Inequality

Suppose  $u, v \in V$ . Then

$$|\langle u, v \rangle| \leq \|u\| \|v\|.$$

This inequality is an equality if and only if one of  $u, v$  is a scalar multiple of the other.

*Proof.* Consider two cases:

*Case 1.*  $v = 0$  and in this case,  $\langle u, v \rangle = 0$ ,  $\|u\| \cdot \|v\| = \|u\| \cdot 0 = 0$ . So the inequality holds.

*Case 2.*  $v \neq 0$ . Consider the orthogonal decomposition

$$u = cv + w$$

where  $c = \frac{\langle u, v \rangle}{\langle v, v \rangle}$  and  $w = u - cv$ . We know that  $w \perp v$ . By Pythagoras' Theorem,

$$\begin{aligned} \|u\|^2 &= \|cv\|^2 + \|w\|^2 \\ &\geq \|cv\|^2 \\ &= |c|^2 \|v\|^2 \\ &= \left| \frac{\langle u, v \rangle}{\|v\|^2} \right|^2 \|v\|^2 \\ &= \frac{|\langle u, v \rangle|^2}{\|v\|^4} \cdot \|v\|^2 \\ &= \frac{|\langle u, v \rangle|^2}{\|v\|^2}. \end{aligned}$$

We just proved that

$$\|u\|^2 \geq \frac{|\langle u, v \rangle|^2}{\|v\|^2}.$$

Multiply both sides of the equation by  $\|v\|^2$  and we get

$$\|u\|^2 \|v\|^2 \geq |\langle u, v \rangle|^2.$$

Take the square root of both sides of the equation and we get

$$\|u\| \cdot \|v\| \geq |\langle u, v \rangle|$$

which is the Cauchy-Schwarz inequality. □

### 1.11.1 Examples of the Cauchy-Schwarz Inequality

(a) If  $x_1, \dots, x_n, y_1, \dots, y_n \in \mathbb{R}$  then

$$|x_1 y_1 + \dots + x_n y_n|^2 \leq (x_1^2 + \dots + x_n^2)(y_1^2 + \dots + y_n^2).$$

(b) If  $f, g$  are continuous real-valued functions on  $[-1, 1]$ , then

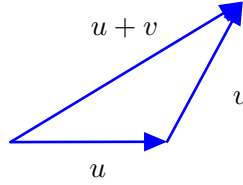
$$\left| \int_{-1}^1 f(x)g(x)dx \right|^2 \leq \left( \int_{-1}^1 (f(x))^2 dx \right) \left( \int_{-1}^1 (g(x))^2 dx \right).$$

### 1.12 Triangle Inequality

The Triangle Inequality implies that the shortest path between two points is a line segment. Suppose  $u, v \in V$ . Then

$$\|u + v\| \leq \|u\| + \|v\|.$$

This inequality is an equality if and only if one of  $u, v$  is a nonnegative multiple of the other.



*Proof.* We have

$$\begin{aligned} \|u + v\|^2 &= \langle u + v, u + v \rangle \\ &= \langle u, u \rangle + \langle u, v \rangle + \langle v, u \rangle + \langle v, v \rangle \\ &= \langle u, u \rangle + \langle u, v \rangle + \overline{\langle u, v \rangle} + \langle v, v \rangle \\ &\leq \langle u, u \rangle + 2|\langle u, v \rangle| + \langle v, v \rangle \\ &\leq \langle u, u \rangle + 2\|u\|\|v\| + \langle v, v \rangle \quad (\text{Cauchy-Schwarz}) \\ &= \|u\|^2 + 2\|u\|\|v\| + \|v\|^2 \\ &= (\|u\| + \|v\|)^2 \end{aligned}$$

Taking the square roots:

$$\|u + v\| \leq \|u\| + \|v\|,$$

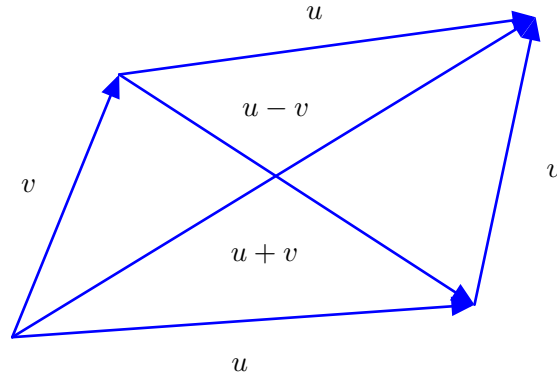
thus we get the triangle inequality. □

### 1.13 Parallelogram Equality

In every parallelogram, the sum of the squares of the lengths of the diagonals equals the sum of the squares of the lengths of the four sides.

Suppose  $u, v \in V$ . Then

$$\|u + v\|^2 + \|u - v\|^2 = 2(\|u\|^2 + \|v\|^2).$$



*Proof.* We have

$$\begin{aligned}\|u + v\|^2 + \|u - v\|^2 &= \langle u + v, u + v \rangle + \langle u - v, u - v \rangle \\ &= \|u\|^2 + \|v\|^2 + \langle u, v \rangle + \langle v, u \rangle \\ &\quad + \|u\|^2 + \|v\|^2 - \langle u, v \rangle - \langle v, u \rangle \\ &= 2(\|u\|^2 + \|v\|^2),\end{aligned}$$

as desired. □

## 2 6.B Orthonormal Bases

## 3 6.C Orthogonal Complements and Minimization Problems