

MATH 311

LAST CHAPTER

SECTION 5.3: ORTHOGONALITY

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Dot Product

If \mathbf{x} is an $1 \times n$ ~~column~~^{row} vector and \mathbf{y} is an $n \times 1$ column vector, then recall that

$$\begin{bmatrix} x_1 & x_2 & \cdots & x_n \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix} = [x_1y_1 + x_2y_2 + \cdots + x_ny_n].$$

The result is a 1×1 matrix that we treat as a number.

DEFINITION 1. Let $\mathbf{x} = [x_1 \ x_2 \ \dots \ x_n]$ and $\mathbf{y} = [y_1 \ y_2 \ \dots \ y_n]$ be two $1 \times n$ row vectors in \mathbb{R}^n . Their **dot product** is defined as followed:

$$\mathbf{x} \cdot \mathbf{y} := \mathbf{xy}^\top = x_1y_1 + x_2y_2 + \cdots + x_ny_n.$$

EXAMPLE 1. If $\mathbf{x} = [1 \ -1 \ -3 \ 1]$ and $\mathbf{y} = [2 \ 1 \ 1 \ 0]$. Then

$$\mathbf{x} \cdot \mathbf{y} = (1)(2) + (-1)(1) + (-3)(1) + (1)(0) = -2.$$

Notes:

- ① We can use other representations of vectors in \mathbb{R}^n .
- ② For instance, if \mathbf{x} and \mathbf{y} are $n \times 1$ column vectors, then

$$\mathbf{x} \cdot \mathbf{y} = x_1y_1 + x_2y_2 + \cdots + x_ny_n = \mathbf{x}^\top \mathbf{y}.$$

Length

DEFINITION 2. Let $\mathbf{x} = [x_1 \ x_2 \ \cdots \ x_n]$. The **length** $\|\mathbf{x}\|$ is defined by

$$\|\mathbf{x}\| := \sqrt{\mathbf{x} \cdot \mathbf{x}} = \sqrt{x_1^2 + x_2^2 + \cdots + x_n^2}.$$

EXAMPLE 2. If $\mathbf{x} = [1 \ 3 \ -2 \ 0]$, then

$$\|\mathbf{x}\| = \sqrt{(1)^2 + (3)^2 + (-2)^2 + (0)^2} = \sqrt{1 + 9 + 4} = \sqrt{14}.$$

Properties:

- ① $\mathbf{x} \cdot \mathbf{y} = \mathbf{y} \cdot \mathbf{x}$.
- ② $\mathbf{x} \cdot (\mathbf{y} + \mathbf{z}) = \mathbf{x} \cdot \mathbf{y} + \mathbf{x} \cdot \mathbf{z}$.
- ③ $(a\mathbf{x}) \cdot \mathbf{y} = a(\mathbf{x} \cdot \mathbf{y}) = \mathbf{x} \cdot (a\mathbf{y})$.
- ④ $\|\mathbf{x}\|^2 = \mathbf{x} \cdot \mathbf{x}$.
- ⑤ $\|\mathbf{x}\| \geq 0$, and $\|\mathbf{x}\| = 0$ if and only if $\mathbf{x} = \mathbf{0}$.
- ⑥ $\|\mathbf{x} + \mathbf{y}\|^2 = \|\mathbf{x}\|^2 + 2(\mathbf{x} \cdot \mathbf{y}) + \|\mathbf{y}\|^2$.

Proof:

$$\begin{aligned} \|\vec{x} + \vec{y}\|^2 &\stackrel{④}{=} (\vec{x} + \vec{y}) \cdot (\vec{x} + \vec{y}) \\ &\stackrel{②}{=} (\vec{x} + \vec{y}) \cdot \vec{x} + (\vec{x} + \vec{y}) \cdot \vec{y} \\ &\stackrel{①}{=} \vec{x} \cdot (\vec{x} + \vec{y}) + \vec{y} \cdot (\vec{x} + \vec{y}) \\ &\stackrel{②}{=} \vec{x} \cdot \vec{x} + \vec{x} \cdot \vec{y} + \vec{y} \cdot \vec{x} + \vec{y} \cdot \vec{y} \end{aligned}$$

$\vec{x} \cdot \vec{y} \stackrel{①}{=}$

Cauchy-Schwarz Inequality

EXAMPLE 3. Let $\mathbf{x} = (a, b)$ and $\mathbf{y} = (c, d)$. Show that

$$|\mathbf{x} \cdot \mathbf{y}| \leq \|\mathbf{x}\| \|\mathbf{y}\|.$$

SOLUTION.

$$|\vec{x} \cdot \vec{y}| \leq \|\vec{x}\| \|\vec{y}\| \Leftrightarrow |\vec{x} \cdot \vec{y}|^2 \leq \|\vec{x}\|^2 \|\vec{y}\|^2$$

So,

$$(ac + bd)^2 \leq (a^2 + b^2)(c^2 + d^2)$$

$$\Leftrightarrow \cancel{a^2 c^2} + 2acbd + \cancel{b^2 d^2} \leq \cancel{a^2 c^2} + a^2 d^2 + b^2 c^2 + \cancel{b^2 d^2}$$

$$\Leftrightarrow 2acbd \leq a^2 d^2 + b^2 c^2$$

$$\Leftrightarrow 0 \leq a^2 d^2 - 2adbc + b^2 c^2$$

$$\Leftrightarrow 0 \leq (ad - bc)^2 \rightarrow \text{this is always true !!!}$$

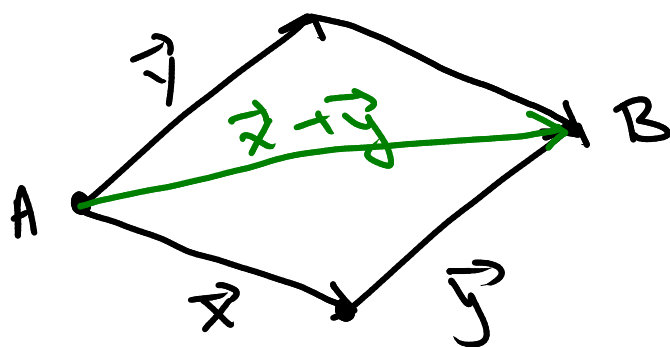
THEOREM 1. If \mathbf{x} and \mathbf{y} are in \mathbb{R}^n , then

$$|\mathbf{x} \cdot \mathbf{y}| \leq \|\mathbf{x}\| \|\mathbf{y}\|.$$

Triangle Inequality

THEOREM 2. If \mathbf{x} and \mathbf{y} are in \mathbb{R}^n , then $\|\mathbf{x} + \mathbf{y}\| \leq \|\mathbf{x}\| + \|\mathbf{y}\|$.

Illustration in \mathbb{R}^2 .



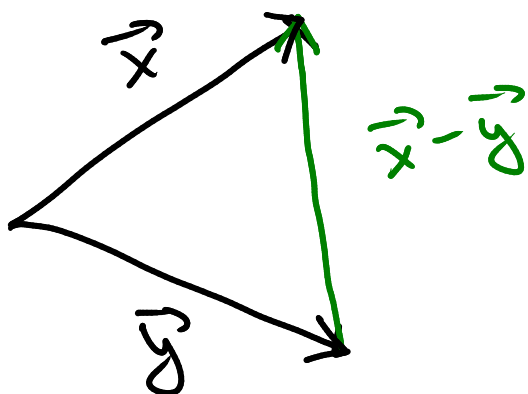
Shortest path
from A to B
is $\|\vec{x} + \vec{y}\|$

Distance

DEFINITION 3. If \mathbf{x} and \mathbf{y} are two vectors in \mathbb{R}^n , the **distance** $d(\mathbf{x}, \mathbf{y})$ is defined by

$$d(\mathbf{x}, \mathbf{y}) = \|\mathbf{x} - \mathbf{y}\|.$$

Illustration in \mathbb{R}^2 .



distance between
 \vec{x} and \vec{y}
= how far the
the tip of \vec{x}
is from \vec{y} .

ORTHOGONALITY

DEFINITION 4. Two vectors \mathbf{x} and \mathbf{y} are **orthogonal** if

$$\mathbf{x} \cdot \mathbf{y} = 0.$$

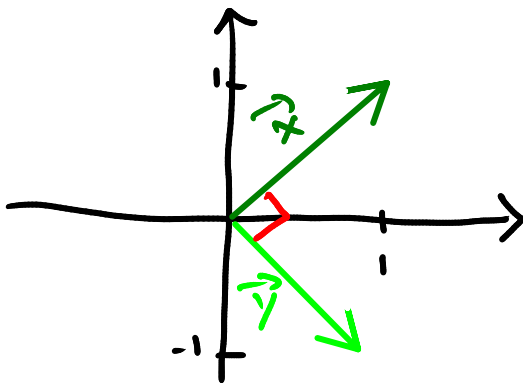
If \mathbf{x} and \mathbf{y} are orthogonal, we write $\mathbf{x} \perp \mathbf{y}$.

EXAMPLE 4. Let $\mathbf{x} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$ and $\mathbf{y} = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$.

- Are \mathbf{x} , \mathbf{y} orthogonal?
- If they are orthogonal, then draw the vectors in a coordinates plane and give one special geometric properties.

$$\begin{aligned} \text{a) } \vec{x} \cdot \vec{y} &= (1)(1) + (1)(-1) = 1 - 1 = 0 \\ &\Rightarrow \vec{x} \perp \vec{y} . \end{aligned}$$

b)



the angle between \vec{x} and \vec{y} is 90° .

Notes: In \mathbb{R}^2 , we can show that

$$\mathbf{x} \cdot \mathbf{y} = \|\mathbf{x}\| \|\mathbf{y}\| \cos \theta$$

where θ is the angle between the vectors \mathbf{x} and \mathbf{y} .

Orthogonal Sets

DEFINITION 5. A collection of vectors $\{\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_k\}$ is an **orthogonal set** if

- ① $\mathbf{x}_i \cdot \mathbf{x}_j = 0$ for any $i \neq j$.
- ② $\mathbf{x}_i \neq 0$ for any i .

EXAMPLE 5. Let

- a) $S_1 = \{(0, 0, 0), (1, 2, 3), (-1, -1, -1)\}$.
- b) $S_2 = \{(1, 2, 3), (-1, -1, -1), (1, 1, 1)\}$.
- c) $S_3 = \{(3, 4, 5), (-4, 3, 0), (-3, -4, 5)\}$.

Which one of these sets is an orthogonal set?

SOLUTION.

a) S_1 is not orth. set because
 $(0, 0, 0) \in S_1$.

b) $(1, 2, 3) \cdot (-1, -1, -1) = -6 \neq 0$
 $\rightarrow S_2$ is not an orth. set.

c) $(3, 4, 5) \cdot (-4, 3, 0) = -12 + 12 + 0 = 0$
 $(3, 4, 5) \cdot (-3, -4, 5) = -9 - 16 + 25 = 0$

$(-4, 3, 0) \cdot (-3, -4, 5) = 12 - 12 = 0$

S_3 is an orth. set.

Orthonormal Sets

DEFINITION 6. A collection of vectors $\{\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_k\}$ is an **orthonormal set** if

- ① it is an orthogonal set.
- ② $\|\mathbf{x}_i\| = 1$ for every index i .

EXAMPLE 6. The standard basis $\{\mathbf{e}_1, \mathbf{e}_2, \dots, \mathbf{e}_n\}$ is an orthonormal set in \mathbb{R}^n .

We can always obtain an orthonormal set from an orthogonal set by **normalizing** the vectors in the orthogonal set.

EXAMPLE 7. Obtain an orthonormal set by normalizing the following orthogonal set:

$$\{(1, -1, 2), (0, 2, 1), (5, 1, -2)\}.$$

SOLUTION. Because $(a\vec{x}) \cdot \vec{y} = a(\vec{x} \cdot \vec{y})$

$$\textcircled{1} \quad a = \|(1, -1, 2)\| = \sqrt{6} \rightarrow \vec{y}_1 = \frac{(1, -1, 2)}{\|(1, -1, 2)\|}$$

$$\Rightarrow \vec{y}_1 = \left(\frac{1}{\sqrt{6}}, -\frac{1}{\sqrt{6}}, \frac{2}{\sqrt{6}} \right)$$

$$\textcircled{2} \quad \vec{y}_2 = \frac{(0, 2, 1)}{\|(0, 2, 1)\|} = \left(0, \frac{2}{\sqrt{5}}, \frac{1}{\sqrt{5}} \right)$$

$$\textcircled{3} \quad \vec{y}_3 = \frac{(5, 1, -2)}{\|(5, 1, -2)\|} = \left(\frac{5}{\sqrt{30}}, \frac{1}{\sqrt{30}}, -\frac{2}{\sqrt{30}} \right)$$

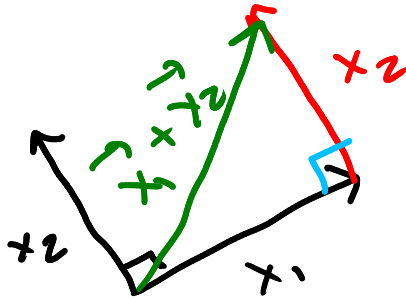
} new
orthonormal
set.

Pythagoras' Theorem

THEOREM 3. If $\{\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_k\}$ is an orthogonal set in \mathbb{R}^n , then

$$\|\mathbf{x}_1 + \mathbf{x}_2 + \dots + \mathbf{x}_k\|^2 = \|\mathbf{x}_1\|^2 + \|\mathbf{x}_2\|^2 + \dots + \|\mathbf{x}_k\|^2.$$

Illustration in \mathbb{R}^2 .



Linearly Independent

THEOREM 4. If $S = \{\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_k\}$ is an orthogonal set in \mathbb{R}^n , then S is linearly independent.

Proof: Let $\vec{x} = a_1 \vec{x}_1 + a_2 \vec{x}_2 + \dots + a_k \vec{x}_k = \vec{0}$
 Then $\vec{x}_1 \cdot \vec{x} = a_1 \vec{x}_1 \cdot \vec{x}_1 + a_2 \vec{x}_1 \cdot \vec{x}_2 + \dots + a_k \vec{x}_1 \cdot \vec{x}_k$
 $\quad \quad \quad = a_1 \|\vec{x}_1\|^2$
 $\Rightarrow 0 = \vec{x}_1 \cdot \vec{x} = a_1 \|\vec{x}_1\|^2$
 $\Rightarrow 0 = a_1$. In this way, $a_2 = a_3 = \dots = a_k = 0$. \square

Fourier Expansion

EXAMPLE 8. Let $U = \text{span}\{(1, -2, 3), (-1, 1, 1)\}$ and $\mathbf{x} = (13, -20, 15) \in U$.

- a) Show $\{(1, -2, 3), (-1, 1, 1)\}$ is an orthogonal basis of U .
- b) Express \mathbf{x} as a linear combination of the basis of U .

SOLUTION.

$$a) (1, -2, 3) \cdot (-1, 1, 1) = -1 - 2 + 3 = 0$$

so, $\{(1, -2, 3), (-1, 1, 1)\}$ is an orthogonal set

$\Rightarrow \{(1, -2, 3), (-1, 1, 1)\}$ is linearly independent.

Because U is spanned by these vectors

$\Rightarrow \{(1, -2, 3), (-1, 1, 1)\}$ is an orthogonal basis for U .

$$(b) \text{ Let } \vec{x} = (13, -20, 15) \\ = a \vec{x}_1 + b \vec{x}_2$$

$$\text{where } \vec{x}_1 = (1, -2, 3), \vec{x}_2 = (-1, 1, 1).$$

First:

$$\vec{x} \cdot \vec{x}_1 = a \vec{x}_1 \cdot \vec{x}_1 = a \|\vec{x}_1\|^2$$

$$\Rightarrow a = \frac{\vec{x} \cdot \vec{x}_1}{\|\vec{x}_1\|^2}$$

$$= \frac{1}{14} \left((13, -20, 15) \cdot (1, -2, 3) \right)$$

$$= \frac{98}{14} = 7$$

Second:

$$\vec{x} \cdot \vec{x}_2 = b \|\vec{x}_2\|^2 \Rightarrow b = \frac{\vec{x} \cdot \vec{x}_2}{\|\vec{x}_2\|^2} = \frac{-18}{3} = -6$$

THEOREM 5. Let $\{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_m\}$ be an orthogonal basis of a subspace U of \mathbb{R}^n . For any $\mathbf{x} \in U$, we have

$$\mathbf{x} = \left(\frac{\mathbf{x} \cdot \mathbf{u}_1}{\|\mathbf{u}_1\|^2} \right) \mathbf{u}_1 + \left(\frac{\mathbf{x} \cdot \mathbf{u}_2}{\|\mathbf{u}_2\|^2} \right) \mathbf{u}_2 + \dots + \left(\frac{\mathbf{x} \cdot \mathbf{u}_m}{\|\mathbf{u}_m\|^2} \right) \mathbf{u}_m.$$

Criteria to be in the Span

EXAMPLE 9. Let $U = \text{span}\{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_m\}$ and let $\mathbf{x} \in \mathbb{R}^n$. Show that if $\mathbf{x} \neq \mathbf{0}$ and $\mathbf{x} \perp \mathbf{u}_k$ for each $1 \leq k \leq m$, then $\mathbf{x} \notin U$.

SOLUTION.

By contradiction, assume $\vec{x} \in U$.

This means by Thm 5

$$\vec{x} = \frac{\vec{x} \cdot \vec{u}_1}{\|\vec{u}_1\|^2} \vec{u}_1 + \frac{\vec{x} \cdot \vec{u}_2}{\|\vec{u}_2\|^2} \vec{u}_2 + \dots + \frac{\vec{x} \cdot \vec{u}_m}{\|\vec{u}_m\|^2} \vec{u}_m.$$

But, $\vec{x} \cdot \vec{u}_k = 0$, for any k

$$\Rightarrow \vec{x} = \frac{0}{\|\vec{u}_1\|^2} \vec{u}_1 + \frac{0}{\|\vec{u}_2\|^2} \vec{u}_2 + \dots + \frac{0}{\|\vec{u}_m\|^2} \vec{u}_m$$

$$= 0\vec{u}_1 + 0\vec{u}_2 + \dots + 0\vec{u}_m$$

$$= \vec{0}.$$

So, a contradiction. Hence $\vec{x} \notin U$. \square