

# *Linear Algebra*



Instructor: Jing YAO

## 3

# Orthogonality (正交性)

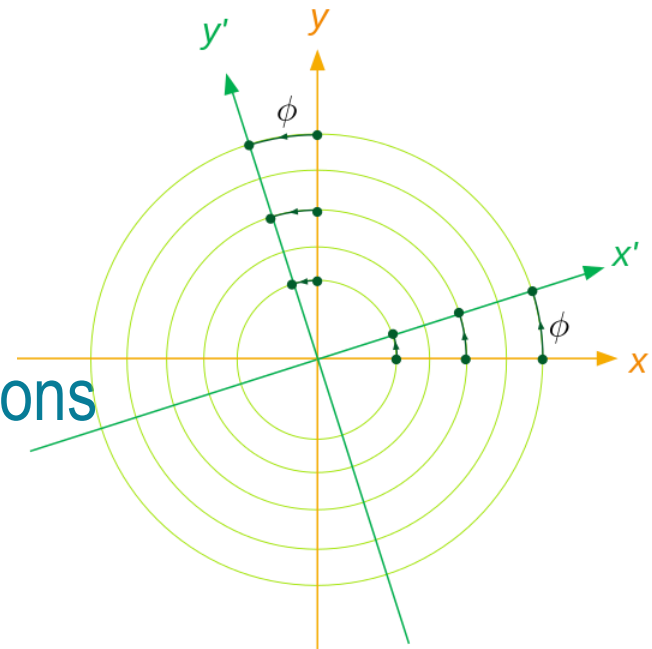
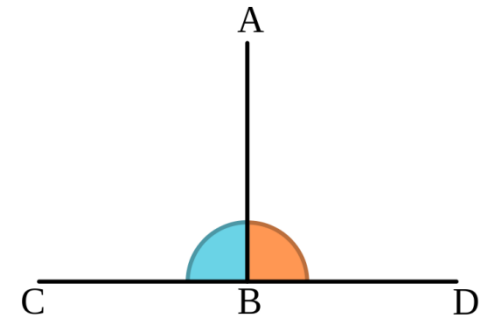
## 3.2

## COSINES AND PROJECTIONS ONTO LINES (余弦; 向量往线上的投影)

Cosines

Projection onto a line

Projections as linear transformations

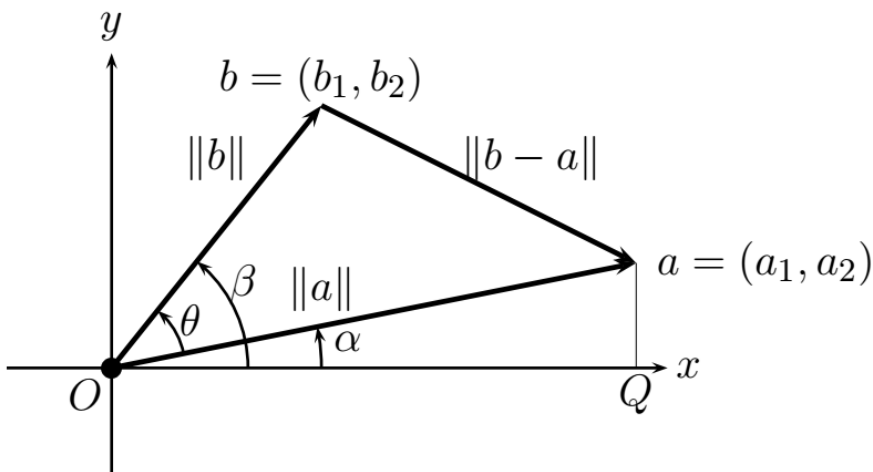
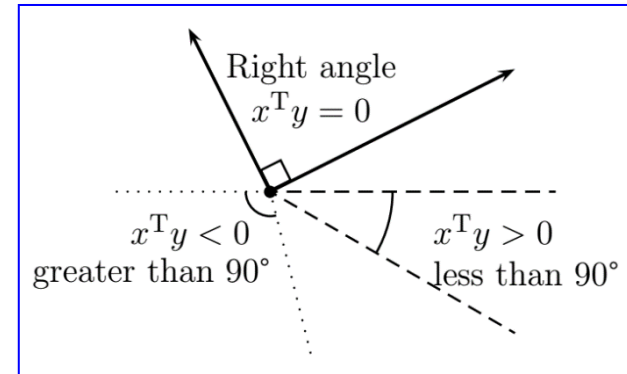


# I. Cosines (余弦)

- If  $\mathbf{x}^T \mathbf{y} = 0$ , then  $\mathbf{x}, \mathbf{y}$  are **orthogonal**, also called **perpendicular**.

*The orthogonal case is the most important.*

Now we allow inner products that are **not zero**, and angles that are **not right angles**.



The cosine of the angle  $\theta = \beta - \alpha$  using inner products.

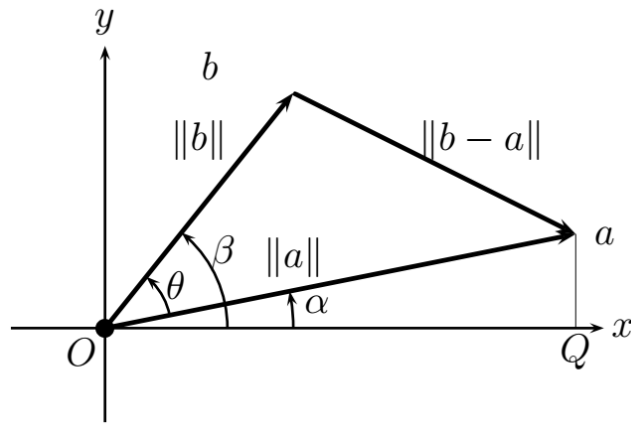
- If  $\mathbf{x}^T \mathbf{y} > 0$ , their angle is less than  $90^\circ$ ;
- If  $\mathbf{x}^T \mathbf{y} < 0$ , their angle is greater than  $90^\circ$ .

$$\theta = \beta - \alpha$$

$$\underline{\cos \theta} = \cos \beta \cos \alpha + \sin \beta \sin \alpha$$

$$= \frac{b_1}{\|\mathbf{b}\|} \frac{a_1}{\|\mathbf{a}\|} + \frac{b_2}{\|\mathbf{b}\|} \frac{a_2}{\|\mathbf{a}\|}$$

$$= \frac{a_1 b_1 + a_2 b_2}{\|\mathbf{a}\| \|\mathbf{b}\|} = \underline{\underline{\frac{\mathbf{a}^T \mathbf{b}}{\|\mathbf{a}\| \|\mathbf{b}\|}}}$$



## Law of Cosines

$$\|b - a\|^2 = \|b\|^2 + \|a\|^2 - 2\|b\|\|a\|\cos \theta$$

$$\Rightarrow (b - a)^T (b - a) = b^T b + a^T a - 2\|b\|\|a\|\cos \theta$$

$$\Rightarrow b^T b - 2a^T b + a^T a = b^T b + a^T a - 2\|b\|\|a\|\cos \theta$$

$$\Rightarrow \cos \theta = \frac{a^T b}{\|a\|\|b\|}.$$

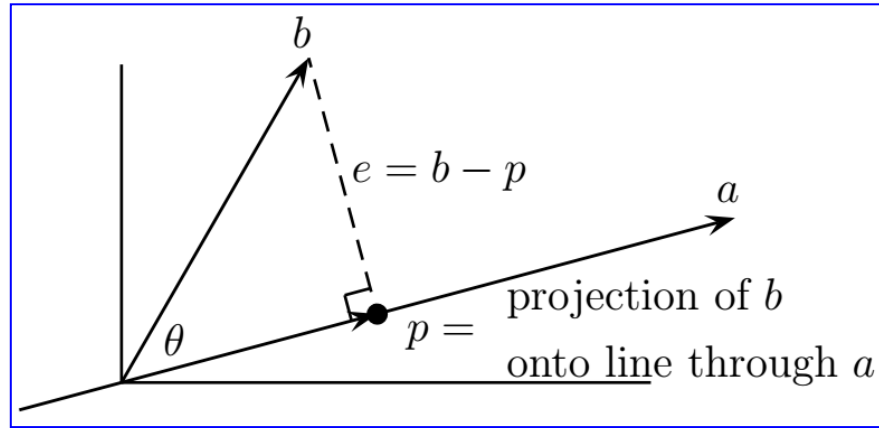
(the cosine of the angle between any *nonzero* vectors  $a$  and  $b$ )

It holds in  $n$  dimensions.

We notice that, since  $|\cos \theta| \leq 1$ , we have

$$\boxed{|a^T b| \leq \|a\|\|b\|}. \quad (\text{Cauchy-Schwarz inequality})$$

## II. Projection onto a Line (往线上的投影)



**Goal:** find the distance from a point  $b$  to the line in the direction of the vector  $a$ .

→ *find the projection  $p$*

(The line connecting  $b$  to  $p$  is perpendicular to  $a$ )

Even though  $a$  and  $b$  are not orthogonal, the distance problem automatically brings in orthogonality.

Let  $\mathbf{a}$  be a vector in a vector space  $V$ , and let  $\text{proj}_{\mathbf{a}}$  be the projection of the vectors of  $V$  onto the line in the direction of  $\mathbf{a}$ . Then

$$\text{proj}_{\mathbf{a}} : \mathbf{b} \mapsto \mathbf{p} = \hat{x}\mathbf{a}$$

for some *scalar*  $\hat{x}$ , and the difference  $\mathbf{b} - \hat{x}\mathbf{a}$  is perpendicular to the vector  $\mathbf{a}$ . Thus

$$0 = \mathbf{a}^T (\mathbf{b} - \hat{x}\mathbf{a}) = \mathbf{a}^T \mathbf{b} - \hat{x} \mathbf{a}^T \mathbf{a},$$

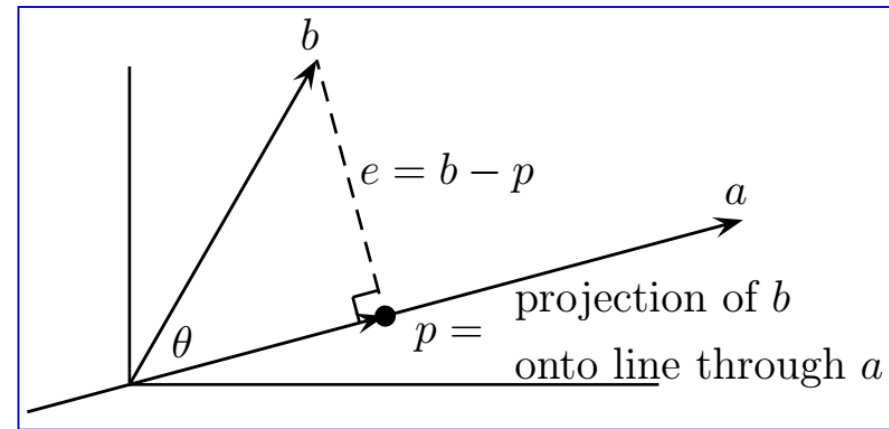
so that the scalar

$$\hat{x} = \frac{\mathbf{a}^T \mathbf{b}}{\mathbf{a}^T \mathbf{a}}.$$

We therefore have the following result.

**Proposition (命题)** *The projection  $\text{proj}_{\mathbf{a}}$  satisfies*

$$\text{proj}_{\mathbf{a}}(\mathbf{b}) = \frac{\mathbf{a}^T \mathbf{b}}{\mathbf{a}^T \mathbf{a}} \mathbf{a}.$$



**Example 1** Project  $\mathbf{b} = (1, 2, 3)^T$  onto the line through  $\mathbf{a} = (1, 1, 1)^T$  to get :

$$\hat{x} = \frac{\mathbf{a}^T \mathbf{b}}{\mathbf{a}^T \mathbf{a}} = \frac{6}{3} = 2.$$

The projection is

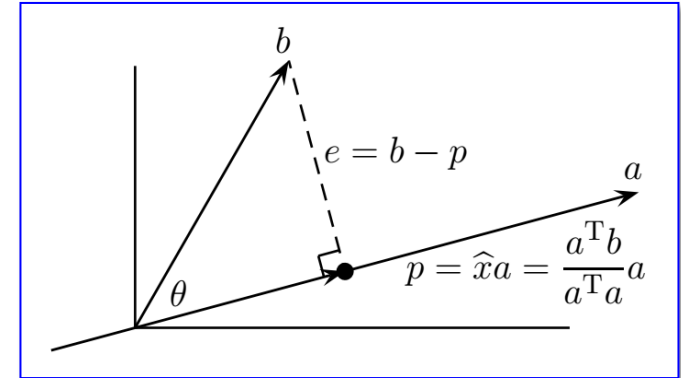
$$\mathbf{p} = \frac{\mathbf{a}^T \mathbf{b}}{\mathbf{a}^T \mathbf{a}} \mathbf{a} = (2, 2, 2)^T.$$

The angle between  $\mathbf{a}$  and  $\mathbf{b}$  has

$$\cos \theta = \frac{\|\mathbf{p}\|}{\|\mathbf{b}\|} = \frac{\sqrt{12}}{\sqrt{14}}.$$

**(Cauchy- Schwarz inequality)**

$$|a^T b| \leq \|a\| \|b\|.$$



**Second proof:**

$$\begin{aligned} \|e\|^2 &= \|b - p\|^2 = \left\| b - \frac{a^T b}{a^T a} a \right\|^2 = \left( b - \frac{a^T b}{a^T a} a \right)^T \left( b - \frac{a^T b}{a^T a} a \right) \\ &= b^T b - 2 \frac{a^T b}{a^T a} a^T b + \left( \frac{a^T b}{a^T a} \right)^2 a^T a \\ &= \frac{(b^T b)(a^T a) - (a^T b)^2}{a^T a} \geq 0 \end{aligned}$$

*Cauchy-Schwarz inequality is equivalent to  $|\cos \theta| \leq 1$ .*

Therefore,  $|a^T b| \leq \|a\| \|b\|.$

*Equality holds if and only if  $b$  is a multiple of  $a$ .*

**Triangle inequality:**  $\|a + b\| \leq \|a\| + \|b\|.$



### III. Projection as a Linear Transformation (Projection Matrix of Rank 1: 秩为1的投影矩阵)

$$\text{proj}_a : \mathbf{b} \mapsto \mathbf{p} = \hat{x}\mathbf{a}.$$

Rewrite the projection  $\text{proj}_a(\mathbf{b})$ :

$$\text{proj}_a(\mathbf{b}) = \frac{\mathbf{a}^T \mathbf{b}}{\mathbf{a}^T \mathbf{a}} \mathbf{a} = \mathbf{a} \frac{\mathbf{a}^T \mathbf{b}}{\mathbf{a}^T \mathbf{a}} = \boxed{\frac{\mathbf{a} \mathbf{a}^T}{\mathbf{a}^T \mathbf{a}}} \mathbf{b}.$$

Projection onto a line is carried out by a *projection matrix*  $\mathbf{P}$ :

$$\mathbf{P} = \frac{\mathbf{a} \mathbf{a}^T}{\mathbf{a}^T \mathbf{a}}. \quad \begin{array}{l} \text{(a column times a row—a square} \\ \text{matrix—divided by the number } \mathbf{a}^T \mathbf{a}.) \end{array}$$

$\mathbf{P}$  is a matrix of rank 1, and as a linear transformation, it transforms a vector  $\mathbf{b}$  to its projection  $\text{proj}_a(\mathbf{b}) = \mathbf{P}\mathbf{b}$ .

**Theorem.** *Let  $\mathbf{a}$  be a nonzero vector of a vector space  $V$ , and let  $T$  be a linear transformation which transforms vector  $\mathbf{b}$  to its projection onto the line in the direction of  $\mathbf{a}$ . Then the matrix of  $T$  is*

$$\mathbf{P} = \frac{\mathbf{a} \mathbf{a}^T}{\mathbf{a}^T \mathbf{a}} = \frac{\mathbf{a} \mathbf{a}^T}{\|\mathbf{a}\|^2}.$$

**Example 2** Let  $\mathbf{a}=(1,1,1)^T$ .

Then the matrix that projects onto the line through  $\mathbf{a}$  is

$$\mathbf{P} = \frac{\mathbf{a}\mathbf{a}^T}{\mathbf{a}^T\mathbf{a}} = \frac{1}{3} \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 \end{bmatrix} = \begin{bmatrix} \frac{1}{3} & \frac{1}{3} & \frac{1}{3} \\ \frac{1}{3} & \frac{1}{3} & \frac{1}{3} \\ \frac{1}{3} & \frac{1}{3} & \frac{1}{3} \end{bmatrix}.$$

This matrix has two properties (typical of projections):

1.  $\mathbf{P}$  is a symmetric matrix:  $\mathbf{P}^T = \mathbf{P}$ .
2. Its square is itself:  $\mathbf{P}^2 = \mathbf{P}$ .

The column space consists of the line through  $\mathbf{a} = (1,1,1)^T$ .

The nullspace consists of the plane perpendicular to  $\mathbf{a}$ .

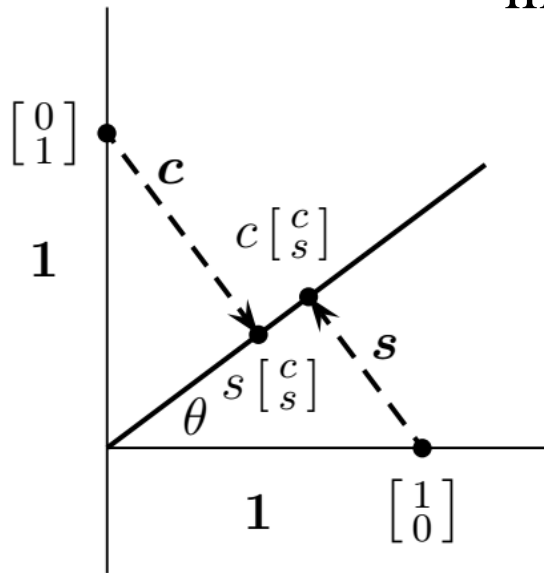
The rank is  $r = 1$ .

**Remark:** The nullspace should be orthogonal to the *row space*.

But because  $\mathbf{P}$  is symmetric, its row and column spaces are the same.

### Example 3 Project onto the “ $\theta$ -direction” in the $x$ - $y$ plane. ( $\mathbf{R}^2$ )

The line goes through  $\mathbf{a} = (\cos\theta, \sin\theta)^T$  and the matrix is symmetric with  $\mathbf{P}^2 = \mathbf{P}$ .



Projection onto the  $\theta$ -line

$$\mathbf{P} = \begin{bmatrix} c^2 & cs \\ cs & s^2 \end{bmatrix}$$

( $c = \cos\theta$ ,  $s = \sin\theta$ )

$$\mathbf{P} = \frac{\mathbf{a}\mathbf{a}^T}{\mathbf{a}^T\mathbf{a}} = \frac{\begin{bmatrix} c \\ s \end{bmatrix} \begin{bmatrix} c & s \end{bmatrix}}{\begin{bmatrix} c & s \end{bmatrix} \begin{bmatrix} c \\ s \end{bmatrix}} = \begin{bmatrix} c^2 & cs \\ cs & s^2 \end{bmatrix}.$$

**Note:**

$\mathbf{P}$  in any number of dimensions:  $\mathbf{P} = \frac{\mathbf{a}\mathbf{a}^T}{\mathbf{a}^T\mathbf{a}}.$

We emphasize that it produces the projection  $\mathbf{p}$ :

To project  $\mathbf{b}$  onto  $\mathbf{a}$ , multiply by the projection matrix  $\mathbf{P}$ :  $\mathbf{p} = \mathbf{P}\mathbf{b}$ .

## Key words:

*Cosine of the angle*

*Projection onto a Line*

*Projection as Linear Transformation: Projection matrix*

## Homework

**See Blackboard**

