

## Lecture #23

### §6.1 Inner Product, Length, Angles, Orthogonality

We shall see now how familiar concepts from  $\mathbb{R}^2$  (plane) or  $\mathbb{R}^3$  (space) can be generalized to an arbitrary  $\mathbb{R}^n$ .

Def: Given two vectors  $\vec{u} = \begin{pmatrix} u_1 \\ \vdots \\ u_n \end{pmatrix}$ ,  $\vec{v} = \begin{pmatrix} v_1 \\ \vdots \\ v_n \end{pmatrix}$  in  $\mathbb{R}^n$ , the inner product (a.k.a. the dot product), denoted  $\vec{u} \cdot \vec{v}$ , is the number  $\vec{u}^T \cdot \vec{v}$

Here:  $\vec{u}^T$  is an  $1 \times n$  matrix,  $\vec{v}$  is an  $n \times 1$  matrix  $\Rightarrow$  product  $\vec{u}^T \cdot \vec{v}$  is a  $1 \times 1$  matrix, i.e. a number.

Ex1: Compute the inner product of  $\vec{u} = \begin{pmatrix} 1 \\ 2 \\ 3 \\ 4 \end{pmatrix}$  and  $\vec{v} = \begin{pmatrix} 5 \\ -3 \\ 1 \\ -2 \end{pmatrix}$  in  $\mathbb{R}^4$ .

$$\vec{u} \cdot \vec{v} = 1 \cdot 5 + 2 \cdot (-3) + 3 \cdot 1 + 4 \cdot (-2) = -6$$

#### Properties of the inner product

1)  $\vec{u} \cdot \vec{v} = \vec{v} \cdot \vec{u}$

2)  $(\vec{u} + \vec{v}) \cdot \vec{w} = \vec{u} \cdot \vec{w} + \vec{v} \cdot \vec{w}$

3)  $(c\vec{u}) \cdot \vec{v} = c(\vec{u} \cdot \vec{v})$

4)  $\vec{u} \cdot \vec{u} \geq 0$  (with equality iff  $\vec{u} = \vec{0}$ )

if  $\vec{u} = \begin{pmatrix} u_1 \\ \vdots \\ u_n \end{pmatrix}$ , then  $\vec{u} \cdot \vec{u} = u_1^2 + u_2^2 + \dots + u_n^2$ .

for any  $\vec{u}, \vec{v}, \vec{w} \in \mathbb{R}^n$   
 $c \in \mathbb{R}$

Def: Given a vector  $\vec{v} = \begin{pmatrix} v_1 \\ \vdots \\ v_n \end{pmatrix}$  in  $\mathbb{R}^n$ , the length (a.k.a. the norm) of  $\vec{v}$ , denoted  $\|\vec{v}\|$ , is defined via  $\|\vec{v}\| := \sqrt{\vec{v} \cdot \vec{v}} = \sqrt{v_1^2 + v_2^2 + \dots + v_n^2} \in \mathbb{R}_{\geq 0}$

Rmk: For  $n=2,3$ , this coincides with the usual notion of length, due to Pythagorean Thm.

Ex2: Find  $\|\vec{u}\|$  and  $\|\vec{v}\|$  in the setup of Ex1

$$\|\vec{u}\| = \sqrt{1^2 + 2^2 + 3^2 + 4^2} = \sqrt{30}$$

$$\|\vec{v}\| = \sqrt{5^2 + (-3)^2 + 1^2 + (-2)^2} = \sqrt{35}$$

## Lecture #23

Properties: 1)  $\|c \cdot \vec{u}\| = \|\vec{u}\| \cdot \underbrace{|c|}_{\text{absolute value of } c}$ .

2)  $\|\vec{u} + \vec{v}\| \leq \|\vec{u}\| + \|\vec{v}\|$  (the triangle inequality)

Def: A vector  $\vec{v}$  is called a unit vector iff  $\|\vec{v}\| = 1$ .

Given a nonzero vector  $\vec{v} \in \mathbb{R}^n$ , a unit vector in the direction of  $\vec{v}$  is given by  $\vec{u} := \frac{1}{\|\vec{v}\|} \cdot \vec{v}$ , while a unit vector in the opposite direction is  $\vec{w} = \frac{-1}{\|\vec{v}\|} \cdot \vec{v}$ .

Ex 3: Find a unit vector in the direction of  $\vec{v} = \begin{pmatrix} 3 \\ 4 \end{pmatrix} \in \mathbb{R}^2$

$\vec{u} = \begin{pmatrix} 3/5 \\ 4/5 \end{pmatrix}$

Def: Given a pair of vectors  $\vec{u}$  and  $\vec{v}$  in  $\mathbb{R}^n$ , the distance b/w  $\vec{u}$  and  $\vec{v}$ , denoted  $\text{dist}(\vec{u}, \vec{v})$ , is the length of  $\vec{u} - \vec{v}$ , i.e.  
 $\text{dist}(\vec{u}, \vec{v}) = \|\vec{u} - \vec{v}\|$

Ex 4: Compute  $\text{dist}(\vec{u}, \vec{v})$  in the setup of Ex 1.

$\text{dist}(\vec{u}, \vec{v}) = \sqrt{(-4)^2 + 5^2 + 2^2 + 6^2} = \sqrt{81} = 9$

Def: Two vectors  $\vec{u}$  and  $\vec{v}$  in  $\mathbb{R}^n$  are orthogonal iff  $\vec{u} \cdot \vec{v} = 0$

Claim: For two vectors  $\vec{u}, \vec{v} \in \mathbb{R}^n$ , we have  
 $\vec{u} \cdot \vec{v} = \|\vec{u}\| \cdot \|\vec{v}\| \cdot \cos \theta$ , where  $\theta$  - angle b/w  $\vec{u}$  and  $\vec{v}$

In particular, two nonzero vectors in  $\mathbb{R}^n$  are orthogonal iff the angle b/w them is  $90^\circ = \pi/2$

Note:  $\vec{0}$  is orthogonal to any  $\vec{u} \in \mathbb{R}^n$ .

## Lecture #23

Claim: Two vectors  $\vec{u}, \vec{v} \in \mathbb{R}^n$  are orthogonal iff  $\|\vec{u} + \vec{v}\|^2 = \|\vec{u}\|^2 + \|\vec{v}\|^2$

It's instructive to present a proof of this claim:

$$\|\vec{u} + \vec{v}\|^2 = (\vec{u} + \vec{v}) \cdot (\vec{u} + \vec{v}) = \underbrace{\vec{u} \cdot \vec{u}}_{\|\vec{u}\|^2} + \underbrace{\vec{v} \cdot \vec{v}}_{\|\vec{v}\|^2} + \underbrace{(\vec{u} \cdot \vec{v} + \vec{v} \cdot \vec{u})}_{2\vec{u} \cdot \vec{v}} \Rightarrow \text{claim follows}$$

The formula  $\vec{u} \cdot \vec{v} = \|\vec{u}\| \cdot \|\vec{v}\| \cdot \cos \theta$  ( $\theta$ -angle b/w  $\vec{u}, \vec{v}$ ) is often used to determine the angle between two vectors.

Ex 5: Find the angle  $\theta$  b/w two vectors in Ex 1.

By Ex 1, Ex 2, have:

$$\cos \theta = \frac{\vec{u} \cdot \vec{v}}{\|\vec{u}\| \cdot \|\vec{v}\|} = \frac{-6}{\sqrt{30} \cdot \sqrt{39}} = -\frac{6}{3 \cdot \sqrt{130}} = -\frac{2}{\sqrt{130}}$$

$$\Rightarrow \theta = \cos^{-1}\left(-\frac{2}{\sqrt{130}}\right)$$

Def: For a subspace  $W$  of  $\mathbb{R}^n$ , its orthogonal complement, denoted  $W^\perp$ , is the set of all  $\vec{z} \in \mathbb{R}^n$  which are orthogonal to any  $\vec{w} \in W$ :

$$W^\perp = \{\vec{z} \in \mathbb{R}^n \mid \vec{z} \cdot \vec{w} = 0 \text{ for any } \vec{w} \in W\}$$

Q: If  $W = \{\vec{0}\} \subseteq \mathbb{R}^n$ , what is  $W^\perp$ ?

A:  $W^\perp = \mathbb{R}^n$

Q: If  $W = \mathbb{R}^n$ , what is  $W^\perp$ ?

A:  $W^\perp = \{\vec{0}\}$

Q: If  $W \subseteq \mathbb{R}^2$  is a subspace consisting of all vectors lying on a given line  $l$  through the origin, describe  $W^\perp$ .

A:  $W^\perp$  is a subspace consisting of all vectors lying on an orthogonal line  $l^\perp$  through the origin.

## Lecture #23

Let  $A$  be an  $m \times n$  matrix. Recall the following subspaces (discussed before):

Col  $A$  - subspace of  $\mathbb{R}^m$

Row  $A$  - subspace of  $\mathbb{R}^n$

Nul  $A$  - subspace of  $\mathbb{R}^n$

Claim:  $(\text{Row } A)^\perp = \text{Nul } A$  and  $(\text{Col } A)^\perp = \text{Nul } A^T$

! Discuss a straightforward proof.

### Properties of $W^\perp$

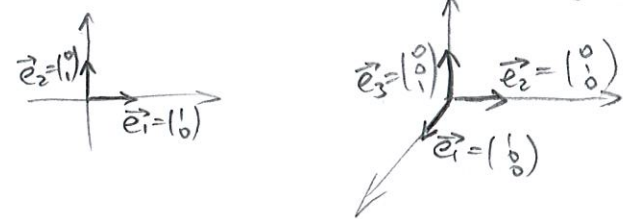
1)  $W^\perp$  is a subspace

2) If  $W$  is spanned by  $\vec{w}_1, \dots, \vec{w}_k \in \mathbb{R}^n$ , then  $W^\perp$  consists of all  $\vec{z} \in \mathbb{R}^n$  orthogonal to each  $\vec{w}_1, \vec{w}_2, \dots, \vec{w}_k$

## §6.2 Orthogonal Sets

Def: A set of vectors  $S = \{\vec{u}_1, \vec{u}_2, \dots, \vec{u}_k\}$  in  $\mathbb{R}^n$  is said to be an orthogonal set if any two distinct vectors in  $S$  are orthogonal:

$$\vec{u}_i \cdot \vec{u}_j = 0 \text{ for any } i \neq j$$

Examples: 

Claim: If  $S = \{\vec{u}_1, \dots, \vec{u}_k\}$  is an orthogonal set of nonzero vectors in  $\mathbb{R}^n$ , then  $S$  is linearly independent.

Again, it is instructive to sketch the proof:

If  $c_1 \vec{u}_1 + c_2 \vec{u}_2 + \dots + c_k \vec{u}_k = \vec{0}$ , then taking inner product with  $\vec{u}_1$

we get  $c_1 (\vec{u}_1 \cdot \vec{u}_1) + 0 + \dots + 0 = 0 \Rightarrow c_1 = 0$  }  $\Rightarrow$  all weights  $= 0$

likewise:  $c_2 = 0, \dots, c_k = 0$

$\Rightarrow S$  - lin. indep.

## Lecture #23

Def: An orthogonal basis for a subspace  $W \subseteq \mathbb{R}^n$  is a basis of  $W$ , which is also an orthogonal set.

Rmk: Due to the previous claim, orthogonal basis for  $W$  is an orthogonal set of elements of  $W$ , spanning  $W$ .

Claim: Let  $\{\vec{u}_1, \vec{u}_2, \dots, \vec{u}_k\}$  be an orthogonal basis for a subspace  $W$  of  $\mathbb{R}^n$ . For each  $\vec{y}$  in  $W$ , the weights in the linear combination

$$\vec{y} = c_1 \vec{u}_1 + c_2 \vec{u}_2 + \dots + c_k \vec{u}_k$$

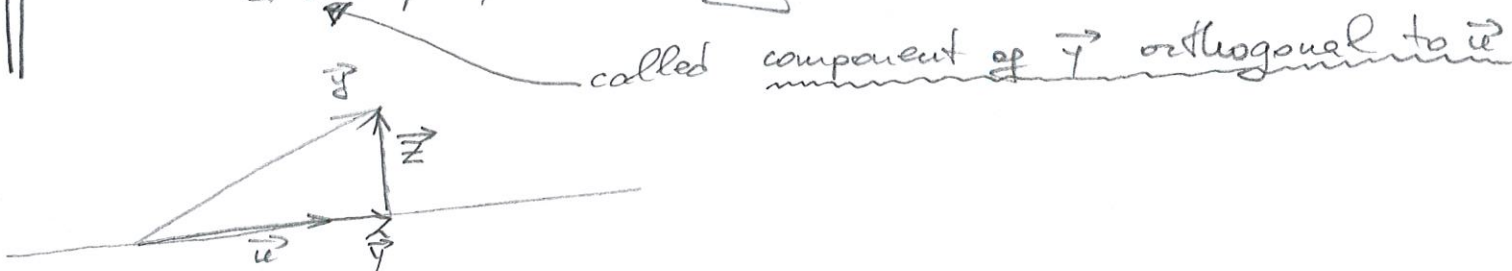
are given by

$$c_j = \frac{\vec{y} \cdot \vec{u}_j}{\vec{u}_j \cdot \vec{u}_j} = \frac{\vec{y} \cdot \vec{u}_j}{\|\vec{u}_j\|^2}$$

Def: Given a nonzero vector  $\vec{u}$  in  $\mathbb{R}^n$ , an orthogonal projection of a vector  $\vec{y} \in \mathbb{R}^n$  onto  $\vec{u}$  is the vector  $\hat{\vec{y}} \in \mathbb{R}^n$  such that

1)  $\hat{\vec{y}}$  is a multiple of  $\vec{u}$

2)  $\vec{z} := \vec{y} - \hat{\vec{y}}$  is orthogonal to  $\vec{u}$



From 1), we see that  $\hat{\vec{y}} = \alpha \cdot \vec{u}$  for some constant  $\alpha$ .

Taking inner product of  $\vec{y} = \alpha \vec{u} + \vec{z}$  with  $\vec{u}$ , we get

$$\vec{y} \cdot \vec{u} = \alpha \cdot \vec{u} \cdot \vec{u} + 0 \Rightarrow \alpha = \frac{\vec{y} \cdot \vec{u}}{\vec{u} \cdot \vec{u}}$$

So: 
$$\hat{\vec{y}} = \frac{\vec{y} \cdot \vec{u}}{\vec{u} \cdot \vec{u}} \cdot \vec{u}$$

Note: The orthogonal projection of  $\vec{y}$  onto  $\vec{u}$  is the same as on any  $c \cdot \vec{u}$ , hence, the notation  $\text{proj}_L \vec{y}$  for  $\hat{\vec{y}}$ , where  $L$ -subspace spanned by  $\vec{u}$ . (5)

## Lecture #23

Ex 6: Verify that  $\vec{u}_1 = \begin{pmatrix} 1 \\ 3 \\ 2 \end{pmatrix}$ ,  $\vec{u}_2 = \begin{pmatrix} 1 \\ -1 \\ 1 \end{pmatrix}$ ,  $\vec{u}_3 = \begin{pmatrix} 5 \\ 1 \\ -4 \end{pmatrix}$  is an orthogonal basis of  $\mathbb{R}^3$

► Verify  $\vec{u}_1 \cdot \vec{u}_2 = 0$ ,  $\vec{u}_1 \cdot \vec{u}_3 = 0$ ,  $\vec{u}_2 \cdot \vec{u}_3 = 0$ .

As they are lin. independent (By Claim) and there 3 of them of height 3, they must be a basis

Def: (1)  $\{\vec{u}_1, \vec{u}_2, \dots, \vec{u}_k\}$  is an orthonormal set of  $\mathbb{R}^n$  iff it is an orthogonal set and each vector is a unit vector

(i.e.  $\vec{u}_i \cdot \vec{u}_j = 0$  for  $i \neq j$  AND  $\vec{u}_i \cdot \vec{u}_i = 1$ )

(2) An orthonormal basis of a subspace  $W$  is a basis of  $W$  that is also an orthonormal set

Ex 7: Construct an orthonormal basis  $\{\vec{v}_1, \vec{v}_2, \vec{v}_3\}$  of  $\mathbb{R}^3$  with  $\vec{v}_i$  in the direction of  $\vec{u}_i$  from Ex 6.

►  $\vec{v}_1 = \frac{1}{\sqrt{14}} \begin{pmatrix} 1 \\ 3 \\ 2 \end{pmatrix}$ ,  $\vec{v}_2 = \frac{1}{\sqrt{3}} \begin{pmatrix} 1 \\ -1 \\ 1 \end{pmatrix}$ ,  $\vec{v}_3 = \frac{1}{\sqrt{42}} \begin{pmatrix} 5 \\ 1 \\ -4 \end{pmatrix}$