## MA-106 Linear Algebra

#### M.K. Keshari

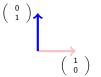


Department of Mathematics Indian Institute of Technology Bombay Powai, Mumbai - 76

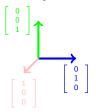
> 27th January, 2015 D1 - Lecture 10

## Orthogonal or perpendicular vectors

The standard basis for  $\mathbb{R}^2$ 



are vectors of length 1 and perpendicular to each other. Similarly the standard basis of  $\mathbb{R}^{3}$ 



are vectors of length 1 and perpendicular to each other.

## Orthogonal vectors

Recall : the length of a vector  $v=(x_1,x_2)^T\in\mathbb{R}^2$  is given by

$$||v|| = \sqrt{x_1^2 + x_2^2} = \sqrt{v^T v}.$$

By Pythagores theorem,  $v,w\in\mathbb{R}^2$  are perpendicular to each other if and only if

$$||v||^{2} + ||w||^{2} = ||v - w||^{2}$$

$$v^{T}v + w^{T}w = (v - w)^{T}(v - w)$$

$$= (v^{T} - w^{T})(v - w)$$

$$= v^{T}v - w^{T}v - v^{T}w + w^{T}w$$

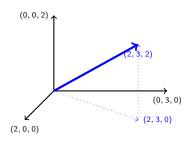
$$v^{T}v + w^{T}w = v^{T}v - 2v^{T}w + w^{T}w \text{ (since } w^{T}v = v^{T}w \text{ )}$$

Therefore, v and w are perpendicular to each other if and only if

$$v^T w = 0.$$

We will generalise this to  $\mathbb{R}^3$  and then to  $\mathbb{R}^n$ .

# length of a vector in $\mathbb{R}^3$ and $\mathbb{R}^n$



Apply Pythagores theorem to the blue triangle to get

$$||(2,3,2)|| = \sqrt{||(2,3,0)||^2 + ||(0,0,2)||^2} = \sqrt{2^2 + 3^2 + 2^2}.$$

Generalize using induction to define the length of a vector  $v = (x_1, x_2, \dots, x_n)^T \in \mathbb{R}^n$  as

$$||v|| = \sqrt{||(x_1, \dots, x_{n-1}, 0)||^2 + ||(0, 0, \dots, x_n)||^2}$$
$$= \sqrt{x_1^2 + \dots + x_{n-1}^2 + x_n^2} = \sqrt{v^T v}.$$

## Inner product and Orthogonal set

Define the **inner product** (dot product) of two vectors  $x, y \in \mathbb{R}^n$  as  $x.y := x^T y$ . Note that

- $x^T y = y^T x = x_1 y_1 + \cdots + x_n y_n$ .
- (Bilinearity)  $(x+y)^T z = x^T z + y^T z$ .  $(cx)^T y = c(x^T y) = x^T (cy)$ .
- $x^T x = ||x||^2 \ge 0$  and  $x^T x = 0$  if and only if x = 0.

#### Definition

As done in  $\mathbb{R}^2$ , two vectors  $v, w \in \mathbb{R}^n$  are orthogonal or perpendicular if and only if  $v^T w = 0$  (consider Span of v, w).

A set of vectors  $\{v_1, v_2, \dots\} \subseteq \mathbb{R}^n$  is an **orthogonal set** if  $v_i \neq 0$  for all i and  $v_i^T v_j = 0$  for all  $i \neq j$ .

An orthogonal set  $\{v_1, v_2, \dots\} \subseteq \mathbb{R}^n$  with  $||v_i|| = 1$  for all i is called an **orthonormal set**.

# Matrices with columns as orthogonal set

$$I = \begin{pmatrix} 1 & & 0 \\ & \ddots & \\ 0 & & 1 \end{pmatrix}.$$

$$A = \begin{pmatrix} 2 & 1 & -2 \\ 2 & -2 & 1 \\ 1 & 2 & 2 \end{pmatrix}.$$

 $\bullet$  Use 1 and -1 only.

- We will show that an orthogonal set is always linearly independent. Hence above matrices are invertible.
- **5** From A and B, we can get matrices with orthonormal columns by replacing each column vector v by v/||v||.

#### Theorem

Every orthogonal set in  $\mathbb{R}^n$  is a linearly independent set.

#### Proof.

Let  $\{v_1, \dots, v_k\}$  be an orthogonal set in  $\mathbb{R}^n$ , i.e.  $v_i \neq 0$  and  $v_i^T v_j = 0$  for  $i \neq j$ .

Assume for some  $a_1,\cdots,a_k\in\mathbb{R}$ ,

$$a_1v_1 + a_2v_2 + \dots + a_kv_k = 0$$

$$\implies (a_1v_1 + a_2v_2 + \dots + a_kv_k)^T v_1 = 0v_1 = 0$$

$$\implies (a_1v_1^T + a_2v_2^T + \dots + a_kv_k^T)v_1 = 0$$

$$\implies a_1v_1^T v_1 + a_2v_2^T v_1 + \dots + a_kv_k^T v_1 = 0$$

$$\implies a_1v_1^T v_1 = 0$$

$$\implies a_1 = 0$$

Similarly, we can show that  $a_2 = \cdots = a_n = 0$ . Hence  $\{v_1, \cdots, v_k\}$  is linearly independent.

- 1. 140020003 AGRIM GUPTA
- 2. 140020019 BANERJEE TANAY
- 140020027 SAUMIL AGARWAL
- 4. 140020041 KANDHARKAR SWAPNIL SANJAYRAO
- 5. 140020062 SATYAM AGNIHOTRI
- 6. 140020074 GURMEET SINGH BEDI
- 7. 140020095 OJASWA GARG
- 8. 140020111 ALLIKANTI TEJASWINI RAVINDER
- 9. 140020124 PUSHPENDRA KUMAR
- 10. 140050003 MAHAJAN HARSHAL RAJESH
- 11. 140050009 UTKARSH GAUTAM
- 12. 140050011 HIMANSHU PAYAL
- 13. 140050040 KANAPARTHI VEERA VENKATA
- 14. 140050046 RAPOLU VISHWANADH
- 15. 140050050 UPPARA RAGHUVEER
- 16. 140050070 SHAIK ABDUL BASITH
- 17. 140050083 K M. MUGILVANNAN

# Orthogonal Subspaces

#### **Definition**

Let V and W be subspaces of  $\mathbb{R}^n$ . We say V and W are orthogonal to each other (notation:  $V \perp W$ ) if every vector in V is orthogonal to every vector in W.

### Example

 $V = \text{Span of } \{v_1, \dots, v_r\}, \ W = \text{Span } \{w_1, \dots, w_t\}. \ \text{If } v_i^T w_j = 0$  for all i, j, then  $V \perp W$ .

Proof. If 
$$v = a_1v_1 + a_2v_2 \in V$$
,  $w = b_1w_1 + b_2w_2 \in W$ , then 
$$v^T w = (a_1v_1^T + a_2v_2^T)(b_1w_1 + b_2w_2)$$
$$= a_1b_1v_1^T w_1 + a_1b_2v_1^T w_2 + a_2b_1v_2^T w_1 + a_2b_2v_2^T w_2$$
$$= 0$$

using bilinearity.

### Example

Let V be yz-plane (screen wall) and W be xz-plane (side wall). Are V and W orthogonal to each other (as subspaces of  $\mathbb{R}^3$ )?

No. The vector  $e_3 = (0,0,1) \neq 0$  lies in V and W both .

**Remark.** If V and W are orthogonal subspaces of  $\mathbb{R}^n$ , then  $V \cap W = (0)$ . It is a necessary condition.

Is it sufficient also? No.

### Example

$$V = \operatorname{Span} \operatorname{of} \begin{pmatrix} 1 \\ 0 \end{pmatrix}, \ W = \operatorname{Span} \operatorname{of} \begin{pmatrix} 1 \\ 1 \end{pmatrix}.$$

Then  $V\cap W=$  (0), but V and W are not orthogonal, since  $\begin{pmatrix} 1 & 0 \end{pmatrix} \begin{pmatrix} 1 \\ 1 \end{pmatrix}=1.$ 

**Example.** 
$$V = \operatorname{Span} \left\{ \begin{pmatrix} 1 \\ 1 \\ 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \\ 1 \\ 0 \end{pmatrix} \right\}, \ W = \operatorname{Span} \left\{ \begin{pmatrix} 1 \\ -1 \\ 1 \\ 0 \end{pmatrix} \right\}.$$

Then V is orthogonal to W.

**Proof.** It is enough to see that both generators of V are orthogonal to the given generator of W.

**Question.** Can we enlarge W to W' such that  $V \perp W'$  and dim  $V + \dim W' = \dim \mathbb{R}^4 = 4$ .

If 
$$A = \begin{pmatrix} 1 & 1 & 0 & 0 \\ 0 & 1 & 1 & 0 \end{pmatrix}$$
, then  $V = \text{row space of } A \text{ and } W \subset N(A)$ .

By Rank-Nullity Theorem: rank(A)+ nullity(A) = 4.

Since rank(A) = 2, we get Nullity(A) = 2.

Therefore if W' = N(A), then  $V \perp W'$  and dim  $V + \dim W' = \dim \mathbb{R}^4$ .

By inspection,  $W' = \text{Span}\{(1, -1, 1, 0)^T, (0, 0, 0, 1)^T\}.$ 

### Theorem (Fundamental Theorem of Orthogonality - 1)

Let A be a  $m \times n$  matrix.

- 1. The row space of A,  $C(A^T)$  is orthogonal to N(A).
- 2. The column space is orthogonal to the left nullspace,  $N(A^T)$ .

**Example.** Assume rank(A) = 1. Then A has 1 pivot. Therefore the column space has dimension 1 and row space has dimension 1.

For 
$$A = \begin{pmatrix} 1 & 2 \\ 2 & 4 \\ 3 & 6 \end{pmatrix}$$
,  $C(A^T) = \operatorname{Span}\begin{pmatrix} 1 \\ 2 \end{pmatrix}$ ,  $N(A) = \operatorname{Span}\begin{pmatrix} -2 \\ 1 \end{pmatrix}$ ,  $C(A) = \operatorname{Span}\begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix}$ ,  $N(A^T) = \operatorname{plane} y_1 + 2y_2 + 3y_3 = 0$ .

$$\dim(C(A^T)) + \dim(N(A)) = 1 + 1 = 2,$$
  
 $\dim(C(A)) + \dim(N(A^T)) = 1 + 2 = 3.$ 

**Observe:** N(A) = set of all vectors othogonal to row space.  $N(A^T) = \text{set of all vectors othogonal to column space.}$