



### 3 Vector Space, Linear Independence

#### 3.1 Vector Space

**Definition.** Vector Space : a nonempty set  $V$  of vectors such that with any two vectors all their linear combinations  $C_1a + c_2b \in V$  ( $c_1, c_2 =$  any real numbers), and all these vectors obey the following rules.

- (a)  $a + b = b + a.$
- (b)  $a + (b + c) = (a + b) + c = a + b + c.$
- (c) There is a unique zero vector such that  $a + 0 = a$  for all  $a.$
- (d) For each  $a$ , there is a unique vector  $-a$  such that  $a + (-a) = 0.$
- (e)  $1a = a.$
- (f)  $c_1(c_2a) = (c_1c_2)a = c_1c_2a.$
- (g)  $c_1(a + b) = c_1a + c_1b.$
- (h)  $(c_1 + c_2)a = c_1a + c_2a.$

**Example 1.**

$\mathbb{R}^n$ , The vector space that consists only of a zero vector, the vector space of all real  $n$  by  $n$  matrices, ...

**Definition.** Subspace of a vector space is a set of vectors (including 0) that satisfies the following. If  $v$  and  $w$  are vectors in the subspace and  $c_1$  is any scalar, then  $v + w, cv$  is in the subspace.

**Example 2.** three-dimensional space  $\mathbb{R}^3$ .

- A plane through (0,0,0) is a subspace of the full vector space  $\mathbb{R}^3$ .
- A line through (0,0,0)
- The single vector (0,0,0)
- The whole space  $\mathbb{R}^3$

**Definition.** The column space of a matrix  $A : C(A)$   
= all linear combinations of the columns  
= span of the columns

**Note**

- The combinations are all possible vectors  $Ax$ .
- To solve  $Ax = b$ ,  $b$  needs to be a combination of the columns.

$$\therefore Ax = b \text{ is solvable iff } b \in C(A)$$

- If  $A$  is  $m$ -by- $n$ , columns belong to  $\mathbb{R}^m$ .

$$\therefore C(A) \text{ is a subspace of } \mathbb{R}^m.$$

**Example 3.**

$$A = \begin{bmatrix} 1 & 0 \\ 4 & 3 \\ 2 & 3 \end{bmatrix} \quad Ax = x_1 \begin{bmatrix} 1 \\ 4 \\ 2 \end{bmatrix} + x_2 \begin{bmatrix} 0 \\ 3 \\ 3 \end{bmatrix}$$

$$C(A) = \text{plane in } \mathbb{R}^3.$$

- $Ax = b$  is solvable when  $b$  is on that plane.
- $\begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$  is in  $C(A)$   $\therefore Ax = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$  solvable.

**Example 4.** In  $\mathbb{R}^2$ ,

$$I = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \quad A = \begin{bmatrix} 1 & 2 \\ 2 & 4 \end{bmatrix} \quad B = \begin{bmatrix} 1 & 2 & 3 \\ 0 & 0 & 4 \end{bmatrix}$$

- $C(I) = \mathbb{R}^2$  ( $Ax = b$  always solvable)
- $C(A) =$  all linear combinations of  $c_1 \begin{bmatrix} 1 \\ 2 \end{bmatrix} + c_2 \begin{bmatrix} 2 \\ 4 \end{bmatrix}$   
 $= c_0 \begin{bmatrix} 1 \\ 2 \end{bmatrix}$  ( $c_0 \in \mathbb{R}$ )  
 $\therefore C(A)$  is only a line.  
( $Ax = b$  solvable only when  $b$  is on the line.)
- $C(B) = \mathbb{R}^2$  (every  $b$  is attainable)  
But  $Bx = b = \begin{bmatrix} 5 \\ 4 \end{bmatrix}$  has multiple solutions.  
( 2 eqns, 3 unknowns)

**Definition.** The row space of a matrix  $A : R(A)$   
= all linear combinations of the vectors  
= span of the rows

- If  $A$  is  $m$ -by- $n$ , rows belong to  $\mathbb{R}^n$ .  
 $\therefore$  The row space is a subspace of  $\mathbb{R}^n$ .

**Note**

- The row space of  $A = C(A^T)$

**Definition.** The null space of a matrix  $A : N(A)$   
= all solutions to  $Ax = 0$

- If  $A$  is  $m$ -by- $n$ , the solution vectors  $x$  are in  $\mathbb{R}^n$ .  
 $\therefore N(A)$  is a subspace of  $\mathbb{R}^n$ .
- Consider  $Ax = b$ .  
If  $b \neq 0$ , then the solutions do not form a subspace.  
( $\because x = 0$  is only a solution if  $b = 0$ .)

**Example 5.**

$$A = [ 1 \quad 2 \quad 3 ]$$

- $N(A)$  is the plane through the origin

$$x + 2y + 3z = 0$$

- $s_1 = \begin{bmatrix} -2 \\ 1 \\ 0 \end{bmatrix}$  and  $s_2 = \begin{bmatrix} -3 \\ 1 \\ 0 \end{bmatrix}$  lie on the plane

$$x + 2y + 3z = 0$$

All vectors on the plane are combinations of  $s_1$  and  $s_2$ .

**Example 6.**

$$A = \begin{bmatrix} 1 & 2 \\ 3 & 8 \end{bmatrix} \implies U = \begin{bmatrix} 1 & 2 \\ 0 & 2 \end{bmatrix} : \text{all columns have pivots.}$$

$$N(A) = \{(0, 0)\}$$

**Example 7.**

$$B = \begin{bmatrix} A \\ 2A \end{bmatrix}$$

Extra rows impose more conditions on the vectors  $x$  in the nullspace, i.e.,  $N(B) = \{(0, 0)\}$

**Example 8.**

$$C = [ A \quad B ] = \begin{bmatrix} 1 & 2 & 2 & 4 \\ 3 & 8 & 6 & 16 \end{bmatrix}$$

Now the solution vector  $x$  has 4 components.

$$\Rightarrow U = \underbrace{\begin{bmatrix} 1 & 2 & 2 & 4 \\ 0 & 2 & 0 & 4 \end{bmatrix}}_{\substack{\text{pivot cols} & \text{free cols}}}$$

$$\Rightarrow R = \begin{bmatrix} 1 & 0 & 2 & 0 \\ 0 & 1 & 0 & 2 \end{bmatrix} \text{ (reduced row echelon form)}$$

Special solutions to  $Rx = 0$  :

$$s_1 = \begin{bmatrix} -2 \\ 0 \\ 1 \\ 0 \end{bmatrix}, \quad s_2 = \begin{bmatrix} 0 \\ -2 \\ 0 \\ 1 \end{bmatrix}$$

( The free variables  $x_3, x_4$  can be given any values whatsoever. Then the pivot variables  $x_1, x_2$  can be found by back substitution.)

All solutions are linear combination of  $s_1$  and  $s_2$ .

**Note**

- If  $A$  has more columns than rows, then  $Ax = 0$  has more unknowns than equations, and it has nonzero solutions. (There must be free columns without pivots.)

$$m \underbrace{\begin{bmatrix} \phantom{x} \\ \phantom{x} \\ \phantom{x} \end{bmatrix}}_n \begin{bmatrix} \phantom{x} \\ \phantom{x} \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

- $N(A) = N(U) = N(R)$

### 3.2 Linear Independence

**Definition.** Vectors  $v_1, \dots, v_n$  are linearly independent.

if.  $x_1v_1 + \dots + x_nv_n = 0$  only happens when all  $x$ 's are zero.

(If a combination is 0 when the  $x$ 's are not all zero, the vectors are dependent.)

- The columns of  $A$  are linear independent when the only solution to  $Ax = 0$  is  $x = 0$ .
- The columns of  $A \in \mathbb{R}^{m \times n}$  are linear independent when the rank is  $r = n$ .
  - $n$  pivots and no free variables.
  - only  $\vec{x} = 0$  is in the nullspace.
- Any set of  $n$  vectors in  $\mathbb{R}^n$  must be linearly dependent if  $n > m$ .

**Definition.** A basis for a vector space is a set of linearly independent vectors that span the space.

- $v_1, \dots, v_n$  are a basis for  $\mathbb{R}^n$  exactly when they are the columns of an  $n \times n$  invertible matrix.

Q. Given  $m$  vectors in  $\mathbb{R}^n$ , how do you find a basis for the space they span?

$$m \begin{bmatrix} - & v_1 & - \\ - & v_2 & - \\ - & \dots & - \\ - & v_n & - \end{bmatrix}$$

$\Downarrow$

eliminate to find the nonzero rows.  
or, put them in columns  $\implies$  find pivot columns.