# Vector Spaces and Subspaces (向量空间和子空间)

Lecture 6

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## **Vector Spaces**

- Vector Space
- 2 Key Examples
- Homework Assignment 6

#### Introduction

#### Recall that:

- The space  $\mathbb{R}^n$  consists of all column vectors with n components.
- $\mathbb{R}^2$  is represented by the usual xy plane.
- The three components of a vector in  $\mathbb{R}^3$  give a point in three-dimensional space.

The valuable thing for linear algebra is that the extension to n dimensions is so straightforward. For a vector in  $\mathbb{R}^7$ , we just need the seven components, even if the geometry is hard to visualize.

## Addition and Scalar Multiplication

Within all vector spaces, we can:

- add any two vectors;
- multiply all vectors by scalars.

In other words, we can take linear combinations.

- Addition obeys the commutative law x+y=y+x; there is a "zero vector" satisfying 0+x=x; and there is a "vector" satisfying -x+x=0.
- Eight properties are fundamental, see the following page.
- A real vector space is a set of vectors together with rules for vector addition and multiplication by real numbers.
- Normally our vectors belong to one of the spaces R<sup>n</sup>; they are
  ordinary vectors. The formal definition allows other things to be
  "vectors"—provided that addition and scalar multiplication are all right.

#### Definition

A real vector space is a set of vectors together with rules for vector addition and multiplication by real numbers. There are eight fundamental properties:

- 1. x+y=y+x (commutativity of addition).
- 2. x + (y+z) = (x+y) + z (associativity of addition).
- 3. There is a unique "zero vector" such that x + 0 = x for all x.
- 4. For each x there is a unique vector -x such that x + (-x) = 0.
- 5. 1x = x.
- 6.  $(c_1c_2)x = c_1(c_2x)$ .
- 7. c(x+y) = cx + cy.
- 8.  $(c_1+c_2)x = c_1x+c_2x$ .

The elements x + y and cx are called the sum of x and y and the product of c and x, respectively.

## **Examples**

Example 1 The infinite-dimensional space  $\mathbb{R}^{\infty}$ . Its vectors have infinitely many components, as in  $x = (1, 2, 1, 2, \cdots)$ . The laws for x + y and cx stay unchanged.

Example 2 The space of 3 by 2 matrices. In this case, the "vectors" are matrices. We can add two matrices, and A+B=B+A, and there is a zero matrix, and so on. This space is almost the same as  $\mathbb{R}^6$ . (The six components are arranged in a rectangle instead of a column.) Any choice of m and n would give, as a similar example, the vector space of all m by n matrices.

## Example 3

Example 3 The space of functions f(x). Here we admit all functions f that are defined on a fixed interval, say  $0 \le x \le 1$ . The space includes  $f(x) = x^2$ ,  $g(x) = \sin x$ , their sum  $(f+g)(x) = x^2 + \sin x$ , and all multiples like  $3x^2$  and  $-\sin x$ . The vectors are functions, and the dimension is somehow a larger infinity than for  $\mathbb{R}^{\infty}$ .

# Additional Properties of Vector Spaces

The following theorem states five more fundamental properties of vector spaces.

#### **Theorem**

If V is a vector space and x, y, and z are elements of V, then

- 1. 0x = 0.
- 2. (-1)x = -x.
- 3. If x + y = x + z, then y = z.
- 4.  $\beta 0 = 0$  for each scalar  $\beta$ .
- 5. If  $\alpha x = 0$ , then either  $\alpha = 0$  or x = 0.

# Example

## Example

Let  $\mathbb{R}^+$  denote the set of positive real numbers. Define the operation of scalar multiplication, denoted  $\circ$ , by

$$\alpha \circ x = x^{\alpha}$$

for each  $x \in \mathbb{R}^+$  and for any real number  $\alpha$ . Define the operation of addition, denoted  $\oplus$ , by

$$x \oplus y = x \cdot y$$
 for all  $x, y \in \mathbb{R}^+$ .

Show that  $\mathbb{R}^+$  is a vector space with these operations.

Thus for this system, the scalar product of -3 times  $\frac{1}{2}$  is given by  $-3 \circ \frac{1}{2} = \left(\frac{1}{2}\right)^{-3} = 8$  and the sum of 2 and 5 is given by  $2 \oplus 5 = 2 \cdot 5 = 10$ .

## Subspace

Consider any plane through the origin in three dimensional space  $\mathbb{R}^3$ , it is a vector space in its own right, and it is actually inside the original space  $\mathbb{R}^3$ . This example suggests us to study *subspace*:

#### Definition

A subspace of a vector space is a nonempty subset that satisfies the requirements for a vector space:

- (i) If we add any vectors x and y in the subspace, x+y is in the subspace.
- (ii) If we multiply any vector x in the subspace by any scalar c, cx is in the subspace.

In other words, Linear combinations stay in the subspace.

#### Remarks

- Notice our emphasis on the word space. A subspace is a subset that is "closed" under addition and scalar multiplication.
- The zero vector belongs to every subspace.
- The distinction between a subset and a subspace: in a subspace, when you add vectors and multiply by scalars, without leaving the space. However, a subset does not generally have that property.

#### Remarks

- The smallest subspace contains only the the zero vector. It is a zero dimensional vector space.
- If the original space is  $\mathbb{R}^3$ , then the possible subspaces are easy to describe:  $\mathbb{R}^3$  itself, any plane through the origin, any line through the origin, or the origin alone.
- The distinction between a subset and a subspace is made clear by examples.

## **Examples**

## Example

Example 4 All vectors in  $\mathbb{R}^2$  whose components are positive or zero. This subset is not a subspace. This subset is the first quadrant of the x-y plane; the coordinates satisfy  $x \geq 0$  and  $y \geq 0$ . It is not a subspace, even though it contains zero and addition does leave us within the subset. Rule (ii) is violated, since if the scalar -1 and the vector  $\begin{bmatrix} 1 & 1 \end{bmatrix}$ , the multiple  $cx = \begin{bmatrix} -1 & -1 \end{bmatrix}$  is in the third quadrant instead of the first.

## Example

Example 5 Lower Triangular Matrices and Symmetric Matrices. Start from the vector space of 3 by 3 matrices. One possible subspace is the set of lower triangular matrices. Another is the set of symmetric matrices.

# The Column Space of A

The incredible thing is that the plane that is spanned by the column vectors of a coefficient matrix of system of linear equations is actually a subspace! It is the column space of A, which is denoted by C(A).

#### Definition

The column space of a matrix A contains all linear combinations of the columns of A. It is a subspace of  $\mathbb{R}^m$ .

#### Theorem and Remarks

#### **Theorem**

The system Ax = b is solvable if and only if the vector b can be expressed as a combination of the columns of A. Then b is in the column space.

- (a) The attainable right-hand sides b are all combinations of the columns of A.
- (b) Geometrical meaning: Ax = b can be solved if and only if b lies in the plane that is spanned by the column vectors.

# The Column Space is a subspace

Consider Ax = b as follows:

$$Ax = \begin{bmatrix} 1 & 0 \\ 5 & 4 \\ 2 & 4 \end{bmatrix} \begin{bmatrix} u \\ v \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix} \begin{pmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \end{pmatrix}$$

We can describe all combinations of the two columns geometrically:

- (a) Ax = b can be solved if and only if b lies in the plane that is spanned by the two column vectors (Figure 2.1). This is the thin set of attainable b.
- (b) If b lies off the plane, then it is not a combination of the two columns. In that case Ax = b has no solution.
- (c) What is important is that this plane is not just a subset of  $\mathbb{R}^3$ ; it is a subspace. It is the column space of A, consisting of all combinations of the columns.

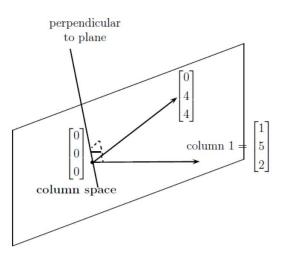


Figure 2.1: The column space C(A), a plane in three-dimensional space.

# Column Space C(A) of an $m \times n$ matrix A is a subspace

It can be readily checked as follows:

- Suppose b and b' lie in the column space, so that Ax = b for some x and Ax' = b' for some x'. Then A(x+x') = b+b', so that b+b' is also a combination of the columns. The column space of all attainable vectors b is closed under addition.
- If b is in the column space C(A), so is any multiple cb. If some combination of columns produces b (say Ax = b), then multiplying that combination by c will produce cb. In other words, A(cx) = cb.

For another matrix A, the dimensions in Figure 2.1 may be very different.

## **Examples**

- zero matrix.
- nonsingular matrix.
- C(A) can be somewhere between the zero space and the whole space. Together with its perpendicular space, it gives one of our two approaches to understand Ax = b.

# The Nullspace of A

#### Definition

The solutions to Ax = 0 form a vector space—the nullspace of A.

The nullspace of a matrix consists of all vectors x such that Ax = 0. It is denoted by N(A). It is subspace of  $\mathbb{R}^n$ , just as the column space was a subspace of  $\mathbb{R}^m$ .

Can you check the requirements (i) and (ii) as well?

## Nullspace: Example

Consider the following example again:

$$Ax = \begin{bmatrix} 1 & 0 \\ 5 & 4 \\ 2 & 4 \end{bmatrix} \begin{bmatrix} u \\ v \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix} \begin{pmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \end{pmatrix}$$

- Only the solutions to a homogeneous equation (b=0) form a subspace.
- The nullspace contains only the vector (0,0).
- The columns of the coefficient matrix are independent.(A key concept that comes soon).

# Nullspace: Another Example

If we add one more column to the previous matrix, which is a combination of the first two columns, then we immediately see that the column space stays the same. However, the Nullspace is quite different.

### Example

If we add an extra column to the previous matrix, what is its new nullspace?

$$\left[\begin{array}{ccc}
1 & 0 & 1 \\
5 & 4 & 9 \\
2 & 4 & 6
\end{array}\right]$$

# Two More Examples

## Example

Let **P** be the plane in 3-space with equation x+2y+z=6. What is the equation of the plane **P**<sub>0</sub> through the origin parallel to **P**? Are **P** and **P**<sub>0</sub> subspaces of  $\mathbb{R}^3$ ?

## Example

If we add an extra column b to a matrix A, then the column space gets larger unless \_\_\_\_\_\_. Give an example in which the column space gets larger and an example in which it doesn't. Why is Ax = b solvable exactly when the column space doesn't get larger by including b?

## Example

### Example

Let  $V = \mathbb{R}^{n \times n}$  be the set of all  $n \times n$  matrices.

- (a) Verify that V is a vector space.
- (b) Let W be the subset of V consisting of all  $n \times n$  matrices such that Tr(A) = 0. Show that W is a subspace of V.
- (c) Can you find the dimension of W?

#### **Further Remarks**

We shall compute the dimensions of those subspaces and a convenient set of vectors to generate them. We hope to end up by understanding all four subspaces that are intimately related to each other and to A-the column space of A, the nullspace of A, and their two perpendicular spaces.

# Homework Assignment 6

2.1: 2, 3, 4, 14, 22, 24, 28.