### 1. Finite dimensional vector spaces

Any finite dimensional vector space can be identified as a Euclidean space.

**Example 1.1.**  $M_{m \times n}(\mathbb{R}) = M_{mn}(\mathbb{R})$ , the space of all real valued  $m \times n$  matrix, can be identified as  $\mathbb{R}^{mn}$ . Every matrix

$$\begin{bmatrix} a_{11} & \cdots & a_{1n} \\ \vdots & \vdots & \vdots \\ a_{m1} & \cdots & a_{mn} \end{bmatrix}$$

is mapped to the column vector

$$\begin{bmatrix} a_{11} & \cdots & a_{1n} & a_{21} & \cdots & a_{2n} & \cdots & a_{m1} & \cdots & a_{mn} \end{bmatrix}^T.$$

**Question:** How to find a basis for  $M_{m \times n}(\mathbb{R})$ ? **Answer:**  $\{M_{ij} : 1 \leq i \leq m, 1 \leq j \leq n\}$  forms a basis for  $M_{m \times n}(\mathbb{R})$ , where  $M_{ij}$  is the  $m \times n$  matrix with 1 in the (i, j)-entry and 0 elsewhere.

**Example 1.2.**  $S_n(\mathbb{R})$ , the space of all real valued symmetric  $n \times n$  matrix, can be identified as  $\mathbb{R}^{\frac{n(n+1)}{2}}$ , thus has dimension  $\frac{n(n+1)}{2}$  (Consider why?). Consider how to find a basis for  $S_n(\mathbb{R})$ . Recall any symmetric matrix is symmetric with respect to the main diagonal.

**Example 1.3.**  $\mathbb{P}_n$ , the space of all polynomials of degree no more than n, can be identified as  $\mathbb{R}^{n+1}$ . Every polynomial

$$a_n x^n + a_{n-1} x^{n-1} + \dots + a_1 x + a_0$$

is mapped to the column vector

$$\begin{bmatrix} a_0 & \cdots & a_n \end{bmatrix}^T$$
.

**Question:** How to find a basis for  $\mathbb{P}_n$ ? **Answer:**  $\{t^n, t^{n-1}, \dots, t, 1\}$  forms a basis for  $\mathbb{P}_n$ .

**Example 1.4.** More generally, any n-dimensional vector space V can be identified as  $\mathbb{R}^n$ . Since V is n-dimensional, we can find a a basis  $\{v_1, \dots, v_n\}$ . For every v in V, we can find a unique linear combination

$$v = c_1 v_1 + \dots + c_n v_n.$$

(Consider why this linear combination is unique). Then the vector v is mapped to the column vector

$$\begin{bmatrix} c_1 & \cdots & c_n \end{bmatrix}^T$$
.

This column vector is called the coordinates of v with respect to the basis  $\{v_1, \dots, v_n\}$ .

# 2. Infinitely dimensional vector spaces

There does exist infinitely dimensional vector space. A vector space is of infinite dimension if it has a basis containing infinitely many vectors.

**Example 2.1.** P := the set of all polynomials is an infinite dimensional vector space.  $\{1, x, x^2, \dots\}$  is a basis of P. This space can be recognized as  $\mathbb{R}^{\infty}$ .

**Example 2.2.** Let I be an interval or the real line  $\mathbb{R}$ .

 $C^n(I) = \{ f : I \to \mathbb{R} : f \ n \ times \ differentiable, \}$ 

 $f, f', \cdots, f^{(n)}$  are all continuous.}

is an infinite dimensional vector space. Indeed, there is a basis of  $C^n(I)$  containing  $\{1, x, x^2, \dots\}$ , and thus has infinitely many elements.

### 3. Subspaces

A subspace S is a subset of a vector space V, which is a vector space itself if equipped with the vector addition and scalar multiplication of V.

**Example 3.1.** State all the subspaces of  $\mathbb{R}^3$ .

**Solution.** Subspaces of  $\mathbb{R}^3$  are  $\mathbb{R}^3$  itself and all the planes and lines passing through the origin.

**Remark 3.2.** A subset S of a vector space V is a subspace if and only if S is closed under the same vector addition and scalar multiplication.

**Remark 3.3.** If  $0_V$  is not in S, then S is not a subspace.

**Example 3.4.** The null space of an  $m \times n$  matrix A, that is, the set of solutions to the homogeneous linear system

$$Ax = 0, \tag{1}$$

is a subspace of  $\mathbb{R}^n$ . The dimension of this subspace is  $n - \operatorname{rank}(A)$ .

**Example 3.5.** The null space of an  $m \times n$  matrix A, that is, the set of solutions to the nonhomogeneous linear system

$$Ax = b \neq 0, \tag{2}$$

is never a subspace of  $\mathbb{R}^n$ .

**Example 3.6.** The set of all the polynomial  $ax^2 + bx + c$  satisfying a + b = c forms a subspace of  $\mathbb{P}_2$ . The dimension of this subspace is 2.(Consider why?)

### 4. LINEAR DEPENDENCE AND INDEPENDENCE

**Proposition 4.1.** A set of vectors  $S = \{v_1, \dots, v_n\}$  is linearly dependent if and only if some  $v_k$  can be expressed as a linear combination of the **other** vectors in S.

**Summary:** How to determine whether a given set  $\{v_1, \dots, v_n\}$  in  $\mathbb{R}^m$  is linearly independent or not?

- If m < n, not linearly independent. (Consider why?)
- If  $m \ge n$ , let  $A = \begin{bmatrix} v_1 & \cdots & v_n \end{bmatrix}$
- If rank(A) = n, that is, the number of columns of A, then  $\{v_1, \dots, v_n\}$  is linearly independent. Otherwise, not.

**Remark 4.2.** When m = n, A is a square matrix. In this case, the last step can be replaced by computing the determinant of A. More precisely, if  $det(A) \neq 0$ , then  $\{v_1, \dots, v_n\}$  is linearly independent. Otherwise, not.

**Summary:** How to find a linearly independent subset out of a given subset  $\{v_1, \dots, v_n\}$  in  $\mathbb{R}^m$ ?

- Let  $A = \begin{bmatrix} v_1 & \cdots & v_n \end{bmatrix}$
- A linearly independent subset of  $\{v_1, \dots, v_n\}$  consists of the columns containing the leading 1's in ref(A).

# 5. Spanning set

**Summary:** How to determine whether a given set  $\{v_1, \dots, v_n\}$  is a spanning set of  $\mathbb{R}^m$ ?

• If m > n, not a spanning set. (Consider why?)

• Let

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$$A = \begin{bmatrix} v_1 & \cdots & v_n \end{bmatrix}.$$

• If rank(A) = the number of rows of A, or equivalently, there is no bottom zero row in ref(A), then  $\{v_1, \dots, v_n\}$  is a spanning set. Otherwise, not.

**Remark 5.1.** When m = n, A is a square matrix. In this case, the last step can be replaced by computing the determinant of A. More precisely, if  $det(A) \neq 0$ , then  $\{v_1, \dots, v_n\}$  is spanning set. Otherwise, not.

Example 5.2. Find a spanning set for the plane

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

in  $\mathbb{R}^3$ .

Solution. This plane actually gives a homogeneous linear system

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

Solving it, we obtain the general expression for the solutions

$$\begin{bmatrix} x \\ y \\ z \end{bmatrix} = s \begin{bmatrix} -2 \\ 1 \\ 0 \end{bmatrix} + t \begin{bmatrix} 3 \\ 0 \\ 1 \end{bmatrix}.$$

Then  $\left\{ \begin{bmatrix} -2\\1\\0 \end{bmatrix}, \begin{bmatrix} 3\\0\\1 \end{bmatrix} \right\}$  is a spanning set of the plane x + 2y - 3z = 0.

L U J L L Remark 5.3. In the above example, the set  $\left\{ \begin{bmatrix} -2\\1\\0 \end{bmatrix}, \begin{bmatrix} 3\\0\\1 \end{bmatrix} \right\}$  is a spanning set of the plane x + 2y - 3z = 0. But the matrix

$$A = \begin{bmatrix} -2 & 3\\ 1 & 0\\ 0 & 1 \end{bmatrix}$$

has rank 2, which is smaller than 3. This seems to be a contradiction. However, in this example, the vector space is not the whole  $\mathbb{R}^3$ , but just a subspace x + 2y - 3z = 0. Recall the argument leading to the conclusion

$$\{v_1, \cdots, v_n\}$$
 spans  $\mathbb{R}^m \iff \operatorname{rank}(A) = m$ .

First, we look at the linear system AX = b for arbitrary  $b \in \mathbb{R}^m$ . If rank(A) < m, then ref(A) has at least one bottom zero row. Since b is arbitrary, we can always find a proper b such that rank(A)  $< \operatorname{rank}(A|b)$ .

But in the above example, b is not arbitrary, since b always belongs to the subspace x + 2y - 3z = 0.

**Theorem 5.4** (rank nullity theorem). Given an  $m \times n$  matrix A, or equivalently a homogeneous linear system Ax = 0, then

 $\operatorname{rank}(A) + \dim(\operatorname{Null space of} A) = n = \text{the number of unknowns.}$ 

**Summary:** How to find a spanning set of a subspace in  $\mathbb{R}^n$ ? (NOT  $\mathbb{R}^n$ !)

- A subspace of  $\mathbb{R}^n$  is usually given by a homogeneous linear system Ax = 0, where A is a  $m \times n$  matrix.
- Solving this linear system, the solutions can be expressed as

$$\vec{x} = c_1 v_1 + \cdots + c_k v_k,$$

where  $c_1, \dots, c_k$  are free parameters, and  $v_1, \dots, v_k$  are fixed vectors in  $\mathbb{R}^n$ . Here

$$k = n - \operatorname{rank}(A)$$

by Theorem 5.4.

• Then  $\{v_1, \dots, v_k\}$  is a spanning set of the subspace.

**Remark 5.5.**  $\{v_1, \dots, v_k\}$  is indeed a basis of this subspace!!

**Theorem 5.6.** If a vector space V is of dimension n and  $\{v_1, \dots, v_n\}$  is a subset of V, then the following statements are equivalent.

- 1.  $\{v_1 \cdots, v_n\}$  is linearly independent.
- 2.  $\{v_1 \cdots, v_n\}$  is a spanning set.
- 3.  $\{v_1 \cdots, v_n\}$  is a basis.

Example 5.7. Is

$$S = \left\{ \begin{bmatrix} 1\\1\\1 \end{bmatrix}, \begin{bmatrix} 3\\0\\1 \end{bmatrix}, \begin{bmatrix} 6\\0\\2 \end{bmatrix} \right\}$$

a spanning set for the plane

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

in  $\mathbb{R}^3$ .

**Solution.** First, it is an easy task to check all three vectors in  $\left\{ \begin{bmatrix} 1\\1\\1 \end{bmatrix}, \begin{bmatrix} 3\\0\\1 \end{bmatrix}, \begin{bmatrix} 6\\0\\2 \end{bmatrix} \right\}$  satisfy x + 2y - 3z = 0. So they all belong to this plane. Let

$$A = \begin{bmatrix} 1 & 3 & 6 \\ 1 & 0 & 0 \\ 1 & 1 & 2 \end{bmatrix} \sim \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 2 \\ 0 & 0 & 0 \end{bmatrix}$$

Therefore,  $\left\{ \begin{bmatrix} 1\\1\\1 \end{bmatrix}, \begin{bmatrix} 3\\0\\1 \end{bmatrix} \right\}$  is linearly independent. Recall by Theorem 5.4, the

dimension of the plane x + 2y - 3z = 0 is 2. By Theorem 5.6,  $\left\{ \begin{bmatrix} 1\\1\\1 \end{bmatrix}, \begin{bmatrix} 3\\0\\1 \end{bmatrix} \right\}$  is a basis of the plane x + 2y - 3z = 0. So S is a spanning set of x + 2y - 3z = 0.

**Summary:** How to determine whether a given set  $\{v_1, \dots, v_k\}$  is a spanning set of a subspace S of  $\mathbb{R}^n$ ? (NOT  $\mathbb{R}^n$ !)

- Find the homogeneous linear system Ax = 0, where A is an  $m \times n$  matrix, representing the subspace S.
- Verify if  $v_i$ 's are solutions to Ax = 0. If one of  $v_i$ 's is not a solution, then this is not a spanning set.
- If all  $v_i$ 's are solutions, then let

 $A = \begin{bmatrix} v_1 & \cdots & v_k \end{bmatrix}.$ 

- Pick up a linearly independent subset out of  $\{v_1, \dots, v_k\}$ . Recall a linearly independent subset of  $\{v_1, \dots, v_k\}$  consists of the columns containing the leading 1's in ref(A).
- Use Theorem 5.4 to find out the dimension of the subspace S.
- If dim S = the number of vectors in the linearly independent subset of  $\{v_1, \dots, v_k\}$ , then  $\{v_1, \dots, v_k\}$  is a spanning set. Otherwise, not.

### 6. Dimensions and bases

**Proposition 6.1.** V is of dimension n. Then

- any linearly independent set cannot contain more than n vectors;
- any spanning set must contain at least n vectors;
- any basis contains exactly n vectors.

**Remark 6.2.** A basis can be considered as a "maximal" linearly independent set, or a "minimal" spanning set.

**Proposition 6.3.** S is a subspace of V. Then  $\dim S \leq \dim V$ . If  $\dim S = \dim V$ , then S = V.

**Summary:** How to determine if a set  $\{v_1, \dots, v_m\}$  is a basis of  $\mathbb{R}^n$ ?

• If  $m \neq n$ , this is not a basis.

- If m = n, let  $A = \begin{bmatrix} v_1 & v_2 & \cdots & v_n \end{bmatrix}$ .
- If  $det(A) \neq 0$ , or equivalently, rank(A) = n = m, this is a basis. Otherwise, it is not.

**Theorem 6.4.** S is a subspace of V. Then any basis of S can be extended to a basis of V.

**Question:** Given a basis of a subspace S, how to extend it to a basis of V?

**Example 6.5.** Extend the basis of span $\begin{pmatrix} -2\\1\\0 \end{pmatrix}$ ,  $\begin{bmatrix} 3\\0\\1 \end{bmatrix}$ ) to a basis of  $\mathbb{R}^3$ .

Solution. We first check

$$\begin{bmatrix} -2 & 3\\ 1 & 0\\ 0 & 1 \end{bmatrix} \sim \begin{bmatrix} 1 & 0\\ 0 & 1\\ 0 & 0 \end{bmatrix}.$$
  
Therefore,  $\{\begin{bmatrix} -2\\ 1\\ 0 \end{bmatrix}, \begin{bmatrix} 3\\ 0\\ 1 \end{bmatrix}\}$  is a basis for span $(\begin{bmatrix} -2\\ 1\\ 0 \end{bmatrix}, \begin{bmatrix} 3\\ 0\\ 1 \end{bmatrix})$ . To find the third  
vector extending  $\{\begin{bmatrix} -2\\ 1\\ 0 \end{bmatrix}, \begin{bmatrix} 3\\ 0\\ 1 \end{bmatrix}\}$  into a basis for  $\mathbb{R}^3$ , we look at  
$$A = \begin{bmatrix} -2 & 3 & 1 & 0 & 0\\ 1 & 0 & 0 & 1 & 0\\ 0 & 1 & 0 & 0 & 1 \end{bmatrix}.$$

The last three column vectors in A is the standard basis for  $\mathbb{R}^3$ . Thus, the five column vectors of A is a spanning set of  $\mathbb{R}^3$ . Moreover,

$$A \sim \begin{bmatrix} 1 & 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 & 1 \\ 0 & 0 & 1 & 2 & 3 \end{bmatrix}.$$

From this observation, we know that the first three columns of A are linearly independent, and thus  $\left\{ \begin{bmatrix} -2\\1\\0 \end{bmatrix}, \begin{bmatrix} 3\\0\\1 \end{bmatrix}, \begin{bmatrix} 1\\0\\0 \end{bmatrix} \right\}$  form a basis for  $\mathbb{R}^3$ .

**Summary:** How to extend a basis for a subspace S to a basis  $\mathbb{R}^n$ ?

- Find a basis  $\{v_1, \cdots, v_k\}$  for S.
- Let  $A = \begin{bmatrix} v_1 & v_2 & \cdots & v_k & u_1 & u_2 & \cdots & u_n \end{bmatrix}$ . Here  $\{u_1, \cdots, u_n\}$  is a basis (usually, we take this set to be the standard basis) of  $\mathbb{R}^n$ .

• A basis of V is the column vectors corresponding to the columns containing the leading 1's in ref(A) (these columns will include  $\{v_1, \dots, v_k\}$ ).

**Summary:** How to find a basis for  $\mathbb{R}^n$  or a subspace S of  $\mathbb{R}^n$ ?

- Find a spanning set  $\{v_1, \cdots, v_n\}$ .
- Let  $A = \begin{bmatrix} v_1 & v_2 & \cdots & v_n \end{bmatrix}$ .
- A basis set, or equivalently a linearly independent subset of  $\{v_1, \dots, v_n\}$ , is the column vectors corresponding to the columns containing the leading 1's in ref(A).

# 7. Relationship between spanning sets, liner independence, and bases

Suppose V is an m-dimensional vector space and  $S = \{v_1, \dots, v_n\}$  is a set of vectors in V. Then

- if n > m, then S is linearly dependent;
- if m > n, then S is not a spanning set;
- if  $n \neq m$ , then S is not a basis.

 $S = \{v_1, \cdots, v_n\}$  is a basis for V means

- S is a "maximal" linearly independent subset of V, i.e., for any  $u \in V, \{v_1, \dots, v_n, u\}$  becomes linearly dependent;
- S is a "minimal" spanning set of V, i.e., after removing any vector  $v_k$  from S, the set  $\{v_1, \dots, v_{k-1}, v_{k+1}, \dots, v_n\}$  is not spanning set anymore.

# 8. RANK, ROW AND COLUMN SPACES

$$A_{m \times n} = \begin{bmatrix} r_1 \\ \vdots \\ r_m \end{bmatrix} = \begin{bmatrix} c_1 & \cdots & c_n \end{bmatrix} \text{ is a } m \times n \text{ matrix.}$$

Note that

$$colspace(A) = rowspace(A^T), \quad colspace(A^T) = rowspace(A).$$

Remark 8.1.

- colspace(A) is a subspace of  $\mathbb{R}^m$ .
- rowspace(A) is a subspace of  $\mathbb{R}^n$ .

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**Remark 8.2.** rank(A) = the dimension of <math>colspace(A) = the dimension of $<math>rowspace(A) = rank(A^T).$ 

**Summary:** How to find a basis for colspace(A)?

• The basis for colspace(A) consists of the columns containing the leading 1's in ref(A).

**Summary:** How to find a basis for rowspace(A)?

- The nonzero rows in ref(A) form a basis for rowspace(A). (Note that these rows are not from the original rows of A.) Or
- The columns containing the leading 1's in ref(A) forms a basis for  $rospace(A^T)$ . Taking transpose of these columns, we obtain a basis for rowspace(A). (These rows are not from the original rows of A.)