# Math 221: LINEAR ALGEBRA

Chapter 5. Vector Space  $\mathbb{R}^n$  §5-1. Subspaces and Spanning

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Subspaces of  $\mathbb{R}^{n}$ 

The null space and the image space

The Eigenspace

Linear Combinations and Spanning Sets

Spanning sets of null(A) and im(A)

# Linear Algebra with Applications Lecture Notes

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# Subspaces of $\mathbb{R}^n$

#### Definitions

- 1.  $\mathbb{R}$  denotes the set of real numbers, and is an example of a set of scalars.
- 2.  $\mathbb{R}^{n}$  is the set of all n-tuples of real numbers, i.e.,

$$\mathbb{R}^n = \left\{ (x_1, x_2, \ldots, x_n) \mid x_i \in \mathbb{R}, 1 \leq i \leq n \right\}.$$

3. The vector space  $\mathbb{R}^n$  consists of the set  $\mathbb{R}^n$  written as column matrices, along with the (matrix) operations of addition and scalar multiplication. Unless stated otherwise,  $\mathbb{R}^n$  means the vector space  $\mathbb{R}^n$ .

#### Remark

 $\mathbb{R}^n$  is a concrete example of the abstract vector space will be studied in the next chapter.

A vectors is denoted by a lower case letter with an arrow written over it; for example,  $\vec{u}$ ,  $\vec{v}$ , and  $\vec{x}$  denote vectors.

Another example: 
$$\vec{\mathbf{u}} = \begin{bmatrix} -2 \\ 3 \\ 0.7 \\ 5 \\ \pi \end{bmatrix}$$
 is a vector in  $\mathbb{R}^5$ , written  $\vec{\mathbf{u}} \in \mathbb{R}^5$ .

To save space on the page, the same vector  $\vec{u}$  may be written instead as a row matrix by taking the transpose of the column:

$$\vec{\mathbf{u}} = \begin{bmatrix} -2, & 3, & 0.7, & 5, & \pi \end{bmatrix}$$

We are interested in nice subsets of  $\mathbb{R}^n$ , defined as follows.

# Definition (Subspaces)

A subset U of  $\mathbb{R}^n$  is a subspace of  $\mathbb{R}^n$  if

S1. The zero vector of  $\mathbb{R}^n$ ,  $\vec{0}_n$ , is in U;

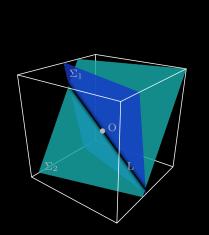
S2. U is closed under addition, i.e., for all  $\vec{u}, \vec{w} \in U, \vec{u} + \vec{w} \in U$ ;

S3. U is closed under scalar multiplication, i.e., for all  $\vec{u}\in U$  and  $k\in \mathbb{R},$   $k\vec{u}\in U.$ 

Both subset  $U = \{\vec{0}_n\}$  and  $R^n$  itself are subspaces of  $\mathbb{R}^n$ . Any other subspace of  $\mathbb{R}^n$  is called a proper subspace of  $\mathbb{R}^n$ .

#### Notation

If U is a subset of  $\mathbb{R}^n,$  we write  $U\subseteq\mathbb{R}^n.$ 



# Example

In  $\mathbb{R}^3$ , the line L through the origin that is parallel to the vector

$$\vec{d} = \begin{bmatrix} -5 \\ 1 \\ -4 \end{bmatrix}$$
 has (vector) equation  $\begin{bmatrix} x \\ y \\ z \end{bmatrix} = t \begin{bmatrix} -5 \\ 1 \\ -4 \end{bmatrix}$ ,  $t \in \mathbb{R}$ , so

$$L = \left\{ t\vec{d} \mid t \in \mathbb{R} \right\}.$$

Claim. L is a subspace of  $\mathbb{R}^3$ .

First: 
$$\vec{0}_3 \in L \text{ since } 0\vec{d} = \vec{0}_3.$$

 $\blacktriangleright$  Suppose  $\vec{u},\vec{v}\in L.$  Then by definition,  $\vec{u}=s\vec{d}$  and  $\vec{v}=t\vec{d},$  for some

$$s,t \in \mathbb{R}$$
. Thus 
$$\vec{u} + \vec{v} = s\vec{d} + t\vec{d} = (s+t)\vec{d}.$$

$$\vec{\mathbf{u}} + \vec{\mathbf{v}} = \mathbf{sd} + \mathbf{td} = (\mathbf{s} + \mathbf{t})\mathbf{d}.$$

Since  $s+t\in\mathbb{R},\ \vec{u}+\vec{v}\in L;\ i.e.,\ L\ is\ closed\ under\ addition.$ 

# Example (continued)

▶ Suppose  $\vec{u} \in L$  and  $k \in \mathbb{R}$  (k is a scalar). Then  $\vec{u} = t\vec{d}$ , for some  $t \in \mathbb{R}$ , so

$$k\vec{u} = k(t\vec{d}) = (kt)\vec{d}.$$

Since  $kt \in \mathbb{R}$ ,  $k\vec{u} \in L$ ; i.e., L is closed under scalar multiplication.

▶ Therefore, L is a subspace of  $\mathbb{R}^3$ .

#### Remark

Note that there is nothing special about the vector  $\vec{d}$  used in this example; the same proof works for any nonzero vector  $\vec{d} \in \mathbb{R}^3$ , so any line through the origin is a subspace of  $\mathbb{R}^3$ .

### Example

In  $\mathbb{R}^3$ , let M denote the plane through the origin having equation

$$3x - 2y + z = 0$$
; then M has normal vector  $\vec{n} = \begin{bmatrix} 3 \\ -2 \\ 1 \end{bmatrix}$ . If  $\vec{u} = \begin{bmatrix} x \\ y \\ z \end{bmatrix}$ , then

$$M = \left\{ \vec{u} \in \mathbb{R}^3 \mid \vec{n} \cdot \vec{u} = 0 \right\},$$

where  $\vec{n} \cdot \vec{u}$  is the dot product of vectors  $\vec{n}$  and  $\vec{u}$ .

Claim. M is a subspace of  $\mathbb{R}^3$ .

- First:  $\vec{0}_3 \in M$  since  $\vec{n} \cdot \vec{0}_3 = 0$ .
- ▶ Suppose  $\vec{u}, \vec{v} \in M$ . Then by definition,  $\vec{n} \cdot \vec{u} = 0$  and  $\vec{n} \cdot \vec{v} = 0$ , so

$$\vec{n} \cdot (\vec{u} + \vec{v}) = \vec{n} \cdot \vec{u} + \vec{n} \cdot \vec{v} = 0 + 0 = 0,$$

and thus  $(\vec{u} + \vec{v}) \in M$ ; i.e., M is closed under addition.

# Example (continued)

▶ Suppose  $\vec{u} \in M$  and  $k \in \mathbb{R}$ . Then  $\vec{n} \cdot \vec{u} = 0$ , so

$$\vec{\mathbf{n}} \cdot (\mathbf{k} \vec{\mathbf{u}}) = \mathbf{k} (\vec{\mathbf{n}} \cdot \vec{\mathbf{u}}) = \mathbf{k} (0) = 0,$$

and thus  $k\vec{u}\in M;$  i.e., M is closed under scalar multiplication.

ightharpoonup Therefore, M is a subspace of  $\mathbb{R}^3$ .

## Remark

As in the previous example, there is nothing special about the plane chosen for this example; any plane through the origin is a subspace of  $\mathbb{R}^3$ .

$$\text{Is } U = \left\{ \begin{bmatrix} a \\ b \\ c \\ d \end{bmatrix} \middle| \begin{array}{l} a,b,c,d \in \mathbb{R} \quad \text{and} \quad 2a-b=c+2d \\ \end{array} \right\} \text{ a subspace of } \mathbb{R}^4?$$
 Justify your answer.

# Solution

The zero vector of 
$$\mathbb{R}^4$$
 is the vector  $\begin{bmatrix} a \\ b \\ c \\ d \end{bmatrix}$  with  $a=b=c=d=0$ .  
In this case,  $2a-b=2(0)+0=0$  and  $c+2d=0+2(0)=0$ , so  $2a-b=c+2d$ . Therefore,  $\vec{0}_4 \in U$ .

# Solution (continued)

Suppose

$$\vec{\mathbf{v}}_1 = \begin{bmatrix} \mathbf{a}_1 \\ \mathbf{b}_1 \\ \mathbf{c}_1 \\ \mathbf{d}_1 \end{bmatrix} \quad \text{and} \quad \vec{\mathbf{v}}_2 = \begin{bmatrix} \mathbf{a}_2 \\ \mathbf{b}_2 \\ \mathbf{c}_2 \\ \mathbf{d}_2 \end{bmatrix} \quad \text{are in U.}$$

Then  $2a_1 - b_1 = c_1 + 2d_1$  and  $2a_2 - b_2 = c_2 + 2d_2$ . Now

$$\vec{v}_1 + \vec{v}_2 = \left[ \begin{array}{c} a_1 \\ b_1 \\ c_1 \\ d_1 \end{array} \right] + \left[ \begin{array}{c} a_2 \\ b_2 \\ c_2 \\ d_2 \end{array} \right] = \left[ \begin{array}{c} a_1 + a_2 \\ b_1 + b_2 \\ c_1 + c_2 \\ d_1 + d_2 \end{array} \right],$$

and

$$\begin{array}{lll} 2(a_1+a_2)-(b_1+b_2) & = & (2a_1-b_1)+(2a_2-b_2) \\ & = & (c_1+2d_1)+(c_2+2d_2) \\ & = & (c_1+c_2)+2(d_1+d_2). \end{array}$$

Therefore,  $\vec{v}_1 + \vec{v}_2 \in U$ .

### Solution (continued)

Finally, suppose

$$ec{
m v} = \left| egin{array}{c} 
m b \\ 
m c \end{array} 
ight| \in {
m U} \quad {
m and} \quad {
m k} \in \mathbb{R}$$

Then 2a - b = c + 2d. Now

$$\mathbf{k}\vec{\mathbf{v}} = \mathbf{k} \begin{bmatrix} \mathbf{a} \\ \mathbf{b} \\ \mathbf{c} \\ \mathbf{d} \end{bmatrix} = \begin{bmatrix} \mathbf{k}\mathbf{a} \\ \mathbf{k}\mathbf{b} \\ \mathbf{k}\mathbf{c} \\ \mathbf{k}\mathbf{d} \end{bmatrix},$$

and

$$2ka - kb = k(2a - b) = k(c + 2d) = kc + 2kd.$$

Therefore,  $k\vec{v} \in U$ .

It follows from the Subspace Test that U is a subspace of  $\mathbb{R}^4$ .

Is  $U = \left\{ \begin{array}{c|c} 1 \\ s \\ t \end{array} \middle| s, t \in \mathbb{R} \right\}$  a subspace of  $\mathbb{R}^3$ ? Justify your answer.

## Solution

Note that  $\vec{0}_3 \notin U$ , and thus U is not a subspace of  $\mathbb{R}^3$ .

(You could also show that U is not closed under addition, or not closed under scalar multiplication.)

Is 
$$U = \left\{ \begin{bmatrix} r \\ 0 \\ s \end{bmatrix} \middle| r, s \in \mathbb{R} \text{ and } r^2 + s^2 = 0 \right\}$$
 a subspace of  $\mathbb{R}^3$ ?

Justify your answer.

# Solution

Since  $r \in \mathbb{R}$ ,  $r^2 \ge 0$  with equality if and only if r = 0. Similarly,  $s \in \mathbb{R}$  implies  $s^2 \ge 0$ , and  $s^2 = 0$  if and only if s = 0. This means  $r^2 + s^2 = 0$  if and only if  $r^2 = s^2 = 0$ ; thus  $r^2 + s^2 = 0$  if and only if r = s = 0. Therefore U contains only  $\vec{0}_3$ , the zero vector, i.e.,  $U = \{\vec{0}_3\}$ . As we already observed,  $\{\vec{0}_n\}$  is a subspace of  $\mathbb{R}^n$ , and therefore U is a subspace of  $\mathbb{R}^3$ .

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Subspaces of  $\mathbb{R}^n$ 

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# The null space and the image space

# Definitions (Null Space and Image Space)

Let A be an  $m \times n$  matrix. The null space of A is defined as

$$\mathrm{null}(A) = \{ \vec{x} \in \mathbb{R}^n \mid A\vec{x} = \vec{0}_m \},$$

and the image space of A is defined as

$$im(A) = \{A\vec{x} \mid \vec{x} \in \mathbb{R}^n\}.$$

#### Remark

- 1. Since A is  $m \times n$  and  $\vec{x} \in \mathbb{R}^n$ ,  $A\vec{x} \in \mathbb{R}^m$ , so  $im(A) \subseteq \mathbb{R}^m$  while  $null(A) \subseteq \mathbb{R}^n$ .
- 2. Image space is also called column space of A, denoted as col(A):

$$col(A) = span(\vec{a}_1, \cdots, \vec{a}_n) = im(A).$$

Prove that if A is an  $m \times n$  matrix, then null(A) is a subspace of  $\mathbb{R}^n$ .

# Proof.

- S1. Since  $A\vec{0}_n = \vec{0}_m$ ,  $\vec{0}_n \in null(A)$ .
- S2. Let  $\vec{x}, \vec{y} \in \text{null}(A)$ . Then  $A\vec{x} = \vec{0}_{\text{m}}$  and  $A\vec{y} = \vec{0}_{\text{m}}$ , so

$$A(\vec{x}+\vec{y}) = A\vec{x} + A\vec{y} = \vec{0}_m + \vec{0}_m = \vec{0}_m, \label{eq:alpha}$$

and thus  $\vec{x} + \vec{y} \in \text{null}(A)$ .

S3. Let  $\vec{x} \in \text{null}(A)$  and  $k \in \mathbb{R}$ . Then  $A\vec{x} = \vec{0}_m$ , so

$$A(k\vec{x}) = k(A\vec{x}) = k\vec{0}_{m} = \vec{0}_{m},$$

and thus  $k\vec{x} \in \text{null}(A)$ .

Therefore, null(A) is a subspace of  $\mathbb{R}^n$ .

Prove that if A is an  $m \times n$  matrix, then im(A) is a subspace of  $\mathbb{R}^m$ .

### Proof.

- S1. Since  $\vec{0}_n \in \mathbb{R}^n$  and  $A\vec{0}_n = \vec{0}_m$ ,  $\vec{0}_m \in im(A)$ .
- S2. Let  $\vec{x}, \vec{y} \in \text{im}(A)$ . Then  $\vec{x} = A\vec{u}$  and  $\vec{y} = A\vec{v}$  for some  $\vec{u}, \vec{v} \in \mathbb{R}^n$ , so

$$\vec{x} + \vec{y} = A\vec{u} + A\vec{v} = A(\vec{u} + \vec{v}).$$

Since  $\vec{u} + \vec{v} \in \mathbb{R}^n$ , it follows that  $\vec{x} + \vec{y} \in \text{im}(A)$ .

S3. Let  $\vec{x} \in \text{im}(A)$  and  $k \in \mathbb{R}$ . Then  $\vec{x} = A\vec{u}$  for some  $\vec{u} \in \mathbb{R}^n$ , and thus

$$k\vec{x} = k(A\vec{u}) = A(k\vec{u}).$$

Since  $k\vec{u} \in \mathbb{R}^n$ , it follows that  $k\vec{x} \in im(A)$ .

Therefore, im(A) is a subspace of  $\mathbb{R}^m$ .

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The null space and the image space

# The Eigenspace

Linear Combinations and Spanning Sets

Spanning sets of null(A) and im(A)

# The Eigenspace

# Definition (Eigenspace)

Let A be an  $n\times n$  matrix and  $\lambda\in\mathbb{R}.$  The eigenspace of A corresponding to  $\lambda$  is the set

$$E_{\lambda}(A) = \{ \vec{x} \in \mathbb{R}^n \mid A\vec{x} = \lambda \vec{x} \} .$$

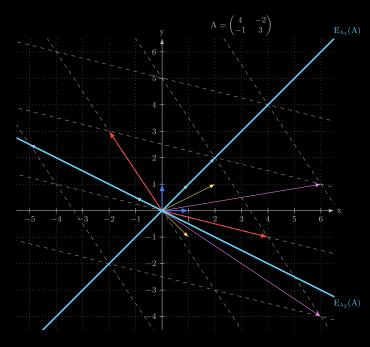
Example

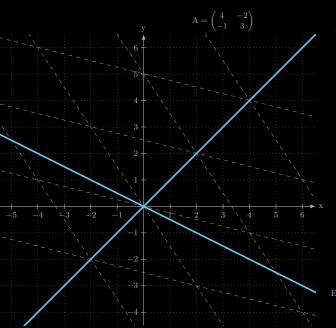
$$A = \begin{pmatrix} 4 & -2 \\ -1 & 3 \end{pmatrix}$$
 has two eigenvalues:  $\lambda_1 = 2$  and  $\lambda_2 = 5$  with corresponding eigenvectors

corresponding eigenvectors 
$$\vec{v}_1 = \begin{pmatrix} 1 \\ 1 \end{pmatrix} \quad \text{and} \quad \vec{v}_2 = \begin{pmatrix} -1 \\ 1/2 \end{pmatrix}$$

Honco

$$egin{aligned} E_{\lambda_1}(A) &= E_2(A) = \{t \vec{v}_1 | t \in \mathbb{R}\} \ E_{\lambda_2}(A) &= E_5(A) = \{t \vec{v}_2 | t \in \mathbb{R}\} \end{aligned}$$





$$\mathbb{E}_2(\mathbf{A}) = \left\{ \mathbf{t} \begin{pmatrix} 1 \\ 1 \end{pmatrix} \middle| \mathbf{t} \in \mathbb{R} \right\}$$

$$\mathrm{E}_5(\mathrm{A}) = \left\{ \mathrm{t} \begin{pmatrix} -1 \\ 1/2 \end{pmatrix} \middle| \mathrm{t} \in \mathbb{R} \right\}$$

Note that

$$\begin{split} E_{\lambda}(A) &= & \left\{ \vec{x} \in \mathbb{R}^n \mid A\vec{x} = \lambda \vec{x} \right\} \\ &= & \left\{ \vec{x} \in \mathbb{R}^n \mid \lambda \vec{x} - A\vec{x} = \vec{0}_n \right\} \\ &= & \left\{ \vec{x} \in \mathbb{R}^n \mid (\lambda I - A)\vec{x} = \vec{0}_n \right\} \end{split}$$

showing that

$$E_{\lambda}(A) = \text{null}(\lambda I - A).$$

It follows that

- ▶ if  $\lambda$  is **not** an eigenvalue of A, then  $E_{\lambda}(A) = \{\vec{0}_n\};$
- ▶ the nonzero vectors of  $E_{\lambda}(A)$  are the eigenvectors of A corresponding to  $\lambda$ ;
- ▶ the eigenspace of A corresponding to  $\lambda$  is a subspace of  $\mathbb{R}^n$ .

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Subspaces of R<sup>n</sup>

The null space and the image space

The Eigenspace

Linear Combinations and Spanning Sets

Spanning sets of null(A) and im(A)

# Linear Combinations and Spanning Sets

# Definition (Linear Combinations and Spanning)

Let  $\vec{x}_1, \vec{x}_2, \dots, \vec{x}_k \in \mathbb{R}^n$  and  $t_1, t_2, \dots, t_k \in \mathbb{R}$ . Then the vector

$$\vec{x} = t_1\vec{x}_1 + t_2\vec{x}_2 + \cdots + t_k\vec{x}_k$$

is called a linear combination of the vectors  $\vec{x}_1, \vec{x}_2, \ldots, \vec{x}_k$ ; the (scalars)  $t_1, t_2, \ldots, t_k \in \mathbb{R}$  are the coefficients. The set of all linear combinations of  $\vec{x}_1, \vec{x}_2, \ldots, \vec{x}_k$  is called the span of  $\vec{x}_1, \vec{x}_2, \ldots, \vec{x}_k$ , and is written

$$\mathrm{span}\{\vec{x}_1, \vec{x}_2, \dots, \vec{x}_k\} = \{t_1\vec{x}_1 + t_2\vec{x}_2 + \dots + t_k\vec{x}_k \mid t_1, t_2, \dots, t_k \in \mathbb{R}\}\,.$$

Additional Terminology. If  $U = \text{span}\{\vec{x}_1, \vec{x}_2, \dots, \vec{x}_k\}$ , then

- ▶ U is spanned by the vectors  $\vec{x}_1, \vec{x}_2, \dots, \vec{x}_k$ .
- ▶ the vectors  $\vec{x}_1, \vec{x}_2, \dots, \vec{x}_k$  span U.
- $\blacktriangleright$  the set of vectors  $\{\vec{x}_1, \vec{x}_2, \dots, \vec{x}_k\}$  is a spanning set for U.

### Example

Let  $\vec{x} \in \mathbb{R}^3$  be a nonzero vector. Then span $\{\vec{x}\} = \{k\vec{x} \mid k \in \mathbb{R}\}$  is a line through the origin having direction vector  $\vec{x}$ .

# Example

Let  $\vec{x},\vec{y}\in\mathbb{R}^3$  be nonzero vectors that are not parallel. Then

$$\mathrm{span}\{\vec{x}, \vec{y}\} = \{k\vec{x} + t\vec{y} \mid k, t \in \mathbb{R}\}$$

is a plane through the origin containing  $\vec{x}$  and  $\vec{y}.$ 

How would you find the equation of this plane?

Let 
$$\vec{\mathbf{x}} = \begin{bmatrix} 8 \\ 3 \\ -13 \\ 20 \end{bmatrix}$$
,  $\vec{\mathbf{y}} = \begin{bmatrix} 2 \\ 1 \\ -3 \\ 5 \end{bmatrix}$  and  $\vec{\mathbf{z}} = \begin{bmatrix} -1 \\ 0 \\ 2 \\ -3 \end{bmatrix}$ . Is  $\vec{\mathbf{x}} \in \operatorname{span}\{\vec{\mathbf{y}}, \vec{\mathbf{z}}\}$ ?

#### Solution

An equivalent question is: can  $\vec{x}$  be expressed as a linear combination of  $\vec{y}$  and  $\vec{z}$ ?

Suppose there exist  $a, b \in \mathbb{R}$  so that  $\vec{x} = a\vec{y} + b\vec{z}$ . Then

$$\begin{bmatrix} 8 \\ 3 \\ -13 \\ 20 \end{bmatrix} = a \begin{bmatrix} 2 \\ 1 \\ -3 \\ 5 \end{bmatrix} + b \begin{bmatrix} -1 \\ 0 \\ 2 \\ -3 \end{bmatrix} = \begin{bmatrix} 2 & -1 \\ 1 & 0 \\ -3 & 2 \\ 5 & -3 \end{bmatrix} \begin{bmatrix} a \\ b \end{bmatrix}.$$

Solve this system of four linear equations in the two variables a and b.

# Solution (continued)

$$\left[\begin{array}{c|c|c} 2 & -1 & 8 \\ 1 & 0 & 3 \\ -3 & 2 & -13 \\ 5 & -3 & 20 \end{array}\right] \rightarrow \left[\begin{array}{c|c|c} 1 & 0 & 3 \\ 0 & 1 & -2 \\ 0 & 0 & 0 \\ 0 & 0 & -1 \end{array}\right]$$

Since the system has no solutions,  $\vec{x} \notin \text{span}\{\vec{y}, \vec{z}\}.$ 

Let 
$$\vec{\mathbf{w}} = \begin{bmatrix} 8 \\ 3 \\ -13 \\ 21 \end{bmatrix}$$
,  $\vec{\mathbf{y}} = \begin{bmatrix} 2 \\ 1 \\ -3 \\ 5 \end{bmatrix}$  and  $\vec{\mathbf{z}} = \begin{bmatrix} -1 \\ 0 \\ 2 \\ -3 \end{bmatrix}$ . Is  $\vec{\mathbf{w}} \in \text{span}\{\vec{\mathbf{y}}, \vec{\mathbf{z}}\}$ ?

This is almost identical to a previous problem, except that  $\vec{w}$  (above) has one entry that is different from the vector  $\vec{x}$  of that problem.

#### Solution

In this case, the system of linear equations is consistent, and gives us  $\vec{w} = 3\vec{y} - 2\vec{z}$ , so  $\vec{w} \in \operatorname{span}\{\vec{y}, \vec{z}\}$ .

#### Theorem

Let  $\vec{x}_1, \vec{x}_2, \dots, \vec{x}_k \in \mathbb{R}^n$  and let  $U = span\{\vec{x}_1, \vec{x}_2, \dots, \vec{x}_k\}$ . Then

- 1. U is a subspace of  $\mathbb{R}^n$  containing each  $\vec{x}_i$ ,  $1 \leq i \leq k$ ;
- 2. if W is a subspace of  $\mathbb{R}^n$  and  $\vec{x}_1, \vec{x}_2, \dots, \vec{x}_k \in W$ , then  $U \subseteq W$ .

# Remark

Property 2 is saying that U is the "smallest" subspace of  $\mathbb{R}^n$  that contains  $\vec{x}_1, \vec{x}_2, \dots, \vec{x}_k$ .

Proof. (Part 1 of Theorem)

Since 
$$U = \text{span}\{\vec{x}_1, \vec{x}_2, \dots, \vec{x}_k\}$$
 and  $0\vec{x}_1 + 0\vec{x}_2 + \dots + 0\vec{x}_k = \vec{0}_n, \vec{0}_n \in U.$ 

Suppose  $\vec{x}, \vec{y} \in U$ . Then for some  $s_i, t_i \in \mathbb{R}, 1 \le i \le k$ ,

$$ec{\mathbf{x}} = \mathbf{s}_1 \vec{\mathbf{x}}_1 + \mathbf{s}_2 \vec{\mathbf{x}}_2 + \dots + \mathbf{s}_k \vec{\mathbf{x}}_k$$

$$ec{\mathbf{y}} = \mathbf{t}_1 \vec{\mathbf{x}}_1 + \mathbf{t}_2 \vec{\mathbf{x}}_2 + \dots + \mathbf{t}_k \vec{\mathbf{x}}_k$$

Thus

$$\begin{split} \vec{x} + \vec{y} &= (s_1 \vec{x}_1 + s_2 \vec{x}_2 + \dots + s_k \vec{x}_k) + (t_1 \vec{x}_1 + t_2 \vec{x}_2 + \dots + t_k \vec{x}_k) \\ &= (s_1 + t_1) \vec{x}_1 + (s_2 + t_2) \vec{x}_2 + \dots + (s_k + t_k) \vec{x}_k. \end{split}$$

Since  $s_i+t_i\in\mathbb{R}$  for all  $1\leq i\leq k,\, \vec{x}+\vec{y}\in U,\, i.e.,\, U$  is closed under addition.

### Proof. (Part 1 of Theorem – continued)

Suppose  $\vec{x} \in U$  and  $a \in \mathbb{R}$ . Then for some  $s_i \in \mathbb{R}$ ,  $1 \le i \le k$ ,

$$\vec{x} = s_1\vec{x}_1 + s_2\vec{x}_2 + \dots + s_k\vec{x}_k$$

Thus

$$\begin{array}{rcl} a\vec{x} & = & a(s_1\vec{x}_1 + s_2\vec{x}_2 + \dots + s_k\vec{x}_k) \\ & = & (as_1)\vec{x}_1 + (as_2)\vec{x}_2 + \dots + (as_k)\vec{x}_k. \end{array}$$

Since  $as_i\in\mathbb{R}$  for all  $1\leq i\leq k,$   $a\vec{x}\in U.$  Hence, U is closed under scalar multiplication.

Therefore, U is a subspace of  $\mathbb{R}^n$ . Furthermore, since

$$\vec{x}_i = \sum_{i=1}^{i-1} 0 \vec{x}_j + 1 \vec{x}_i + \sum_{i=i+1}^{k} 0 \vec{x}_j,$$

it follows that  $\vec{x}_i \in U$  for all  $i, 1 \le i \le k$ .

### Proof. (Part 2 of Theorem)

Let  $W \subset \mathbb{R}^n$  be a subspace that contains  $\vec{x}_1, \dots, \vec{x}_n$ . We need to prove that  $U \subseteq W$ .

Suppose  $\vec{x} \in U$ . Then  $\vec{x} = s_1\vec{x}_1 + s_2\vec{x}_2 + \dots + s_k\vec{x}_k$  for some  $s_i \in \mathbb{R}, 1 \leq i \leq k$ . Since W contain each  $\vec{x}_i$  and W is closed under scalar multiplication, it follows that  $s_i\vec{x}_i \in W$  for each  $i, 1 \leq i \leq k$ . Furthermore, since W is closed under addition,  $\vec{x} = s_1\vec{x}_1 + s_2\vec{x}_2 + \dots + s_k\vec{x}_k \in W$ . Therefore,  $U \subseteq W$ .

Problem (revisited)

Is 
$$U = \left\{ \begin{bmatrix} a \\ b \\ c \\ d \end{bmatrix} \middle| a, b, c, d \in \mathbb{R} \text{ and } 2a - b = c + 2d \right\}$$
 a subspace of  $\mathbb{R}^4$ ?

Justify your answer.

## Solution (Another)

Let 
$$\vec{v} = \begin{bmatrix} a \\ b \\ c \\ d \end{bmatrix} \in U$$
. Since  $2a - b = c + 2d$ ,  $c = 2a - b - 2d$ , and thus

$$U = \left\{ \left[ \begin{array}{c} a \\ b \\ 2a - b - 2d \end{array} \right] \; \middle| \; a, b, d \in \mathbb{R} \right\} = \operatorname{span} \left\{ \left[ \begin{array}{c} 1 \\ 0 \\ 2 \\ 0 \end{array} \right], \left[ \begin{array}{c} 0 \\ 1 \\ -1 \\ 0 \end{array} \right], \left[ \begin{array}{c} 0 \\ 0 \\ -2 \\ 1 \end{array} \right] \right\}.$$

By a previous Theorem, U is a subspace of  $\mathbb{R}^4$ .

Let  $\vec{x}, \vec{y} \in \mathbb{R}^n$ ,  $U_1 = \operatorname{span}\{\vec{x}, \vec{y}\}$ , and  $U_2 = \operatorname{span}\{2\vec{x} - \vec{y}, 2\vec{y} + \vec{x}\}$ . Prove that  $U_1 = U_2$ .

#### Solution

To show that  $U_1 = U_2$ , prove that  $U_1 \subseteq U_2$ , and  $U_2 \subseteq U_1$ . We begin by noting that, by the first part of the previous Theorem,  $U_1$  and  $U_2$  are subspaces of  $\mathbb{R}^n$ .

Since  $2\vec{x} - \vec{y}, 2\vec{y} + \vec{x} \in U_1$ , it follows from the second part of the previous Theorem that span $\{2\vec{x} - \vec{y}, 2\vec{y} + \vec{x}\} \subseteq U_1$ , i.e.,  $U_2 \subseteq U_1$ .

Also, since

$$\vec{x} = \frac{2}{5} (2\vec{x} - \vec{y}) + \frac{1}{5} (2\vec{y} + \vec{x}) ,$$
 
$$\vec{y} = -\frac{1}{5} (2\vec{x} - \vec{y}) + \frac{2}{5} (2\vec{y} + \vec{x}) ,$$

 $\vec{x}, \vec{y} \in U_2$ . Therefore, by the second part of the previous Theorem, span $\{\vec{x}, \vec{y}\} \subseteq U_2$ , i.e.,  $U_1 \subseteq U_2$ . The result now follows.

Show that  $\mathbb{R}^n = span\{\vec{e}_1, \vec{e}_2, \dots, \vec{e}_n\}$ , where  $\vec{e}_j$  denote the  $j^{th}$  column of  $I_n$ .

#### Solution

Let 
$$\vec{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \in \mathbb{R}^n$$
. Then  $\vec{x} = x_1 \vec{e_1} + x_2 \vec{e_2} + \dots + x_n \vec{e_n}$ , where  $x_1, x_2, \dots, x_n \in \mathbb{R}$ . Therefore,  $\vec{x} \in \text{span}\{\vec{e_1}, \vec{e_2}, \dots, \vec{e_n}\}$ , and thus

 $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n \in \mathbb{R}$ . Therefore,  $\mathbf{x} \in \text{span} \{ \mathbf{e}_1, \mathbf{e}_2, \dots, \mathbf{e}_n \}$ .

Conversely, since  $\vec{e}_i \in \mathbb{R}^n$  for each  $i, 1 \leq i \leq n$  (and  $\mathbb{R}^n$  is a vector space), it follows that span $\{\vec{e}_1, \vec{e}_2, \dots, \vec{e}_n\} \subseteq \mathbb{R}^n$ . The equality now follows.

$$\text{Let } \vec{x}_1 = \left[ \begin{array}{c} 1 \\ 1 \\ 1 \\ 1 \end{array} \right], \vec{x}_2 = \left[ \begin{array}{c} 0 \\ 1 \\ 1 \\ 1 \end{array} \right], \vec{x}_3 = \left[ \begin{array}{c} 0 \\ 0 \\ 1 \\ 1 \end{array} \right], \vec{x}_4 = \left[ \begin{array}{c} 0 \\ 0 \\ 0 \\ 1 \end{array} \right].$$

Does  $\{\vec{x}_1, \vec{x}_2, \vec{x}_3, \vec{x}_4\}$  span  $\mathbb{R}^4$ ? (Equivalently, is span $\{\vec{x}_1, \vec{x}_2, \vec{x}_3, \vec{x}_4\} = \mathbb{R}^4$ ?)

#### Solution

To prove span $\{\vec{x}_1, \vec{x}_2, \vec{x}_3, \vec{x}_4\} = \mathbb{R}^4$ , we need to prove two directions:

$$\operatorname{span}\{\vec{x}_1,\vec{x}_2,\vec{x}_3,\vec{x}_4\}\subseteq\mathbb{R}^4\quad \text{and}\quad \mathbb{R}^4\subseteq\operatorname{span}\{\vec{x}_1,\vec{x}_2,\vec{x}_3,\vec{x}_4\}.$$

For the first relation, since  $\vec{x}_1, \vec{x}_2, \vec{x}_3, \vec{x}_4 \in \mathbb{R}^4$  (and  $\mathbb{R}^4$  is a vector space), span $\{\vec{x}_1, \vec{x}_2, \vec{x}_3, \vec{x}_4\} \subseteq \mathbb{R}^4$ .

## Solution (continued)

For the second relation, notice that

$$\vec{e}_1 = \vec{x}_1 - \vec{x}_2$$
 $\vec{e}_2 = \vec{x}_2 - \vec{x}_3$ 
 $\vec{e}_3 = \vec{x}_3 - \vec{x}_4$ 

showing that  $\vec{e}_1, \vec{e}_2, \vec{e}_3, \vec{e}_4 \in \text{span}\{\vec{x}_1, \vec{x}_2, \vec{x}_3, \vec{x}_4\}$ . Therefore, since  $\text{span}\{\vec{x}_1, \vec{x}_2, \vec{x}_3, \vec{x}_4\}$  is a vector space,

$$\mathbb{R}^4=\operatorname{span}\{\vec{e}_1,\vec{e}_2,\vec{e}_3,\vec{e}_4\}\subseteq\operatorname{span}\{\vec{x}_1,\vec{x}_2,\vec{x}_3,\vec{x}_4\},$$

and the equality follows.

$$\operatorname{Let}\, ec{\mathrm{u}}_1 = \left[ egin{array}{c} 1 \\ -1 \\ 1 \\ -1 \end{array} \right], ec{\mathrm{u}}_2 = \left[ egin{array}{c} -1 \\ 1 \\ 1 \end{array} \right], ec{\mathrm{u}}_3 = \left[ egin{array}{c} 1 \\ -1 \\ -1 \end{array} \right], ec{\mathrm{u}}_4 = \left[ egin{array}{c} 1 \\ -1 \\ 1 \end{array} \right]$$

Show that span $\{\vec{\mathbf{u}}_1, \vec{\mathbf{u}}_2, \vec{\mathbf{u}}_3, \vec{\mathbf{u}}_4\} \neq \mathbb{R}^4$ .

### Solution

If you check, you'll find that  $\vec{e}_2$  can not be written as a linear combination of  $\vec{u}_1, \vec{u}_2, \vec{u}_3$ , and  $\vec{u}_4$ .

### Copyright

Subspaces of R<sup>n</sup>

The null space and the image space

The Eigenspace

Linear Combinations and Spanning Sets

Spanning sets of null(A) and im(A)

Spanning sets of null(A) and im(A)

### Lemma

Let A be an  $m\times n$  matrix, and let  $\{\vec{x}_1,\vec{x}_2,\ldots,\vec{x}_k\}$  denote a set of basic solutions to  $A\vec{x}=\vec{0}_m.$  Then

$$\mathrm{null}(A) = \mathrm{span}\{\vec{x}_1, \cdots, \vec{x}_k\}.$$

### Lemma

Let A be an  $m \times n$  matrix with columns  $\vec{c}_1, \vec{c}_2, \dots, \vec{c}_n$ . Then

$$\operatorname{im}(A) = \operatorname{span}\{\vec{c}_1, \vec{c}_2, \dots, \vec{c}_n\}.$$

# Proof. (of null(A) = span $\{\vec{x}_1, \dots, \vec{x}_k\}$ )

"]:" Because  $\vec{x}_i \in null(A)$  for each  $i, \ 1 \leq i \leq k,$  it follows that

$$\mathrm{span}\{\vec{x}_1,\vec{x}_2,\ldots,\vec{x}_k\}\subseteq \mathrm{null}(A).$$

" $\subseteq$ :" Every solution to  $A\vec{x} = \vec{0}_m$  can be expressed as a linear combination of basic solutions, implying that

$$\operatorname{null}(A) \subseteq \operatorname{span}\{\vec{x}_1,\vec{x}_2,\ldots,\vec{x}_k\}.$$

Therefore,  $null(A) = span\{\vec{x}_1, \vec{x}_2, \dots, \vec{x}_k\}.$ 

Proof. (of 
$$im(A) = span\{\vec{c}_1, \vec{c}_2, \dots, \vec{c}_n\}$$
)

" $\subseteq$ :" Suppose  $\vec{y} \in im(A)$ . Then (by definition) there is a vector  $\vec{x} \in \mathbb{R}^n$  so that  $\vec{y} = A\vec{x}$ . Write  $\vec{x} = \begin{bmatrix} x_1 & x_2 & \dots & x_n \end{bmatrix}^T$ . Then

$$ec{y} = Aec{x} = \left[ egin{array}{cccc} ec{c}_1 & ec{c}_2 & \dots & ec{c}_n \end{array} 
ight] \left[ egin{array}{c} x_1 \ x_2 \ dots \ x_n \end{array} 
ight] = x_1 ec{c}_1 + x_2 ec{c}_2 + \dots + x_n ec{c}_n.$$

Therefore,  $\vec{y} \in \text{span}\{\vec{c}_1, \vec{c}_2, \dots, \vec{c}_n\}$ , implying that

im(A) 
$$\subseteq$$
 span $\{ec{c}_1,ec{c}_2,\ldots,ec{c}_n\}$ , implying that

## Proof. (continued)

Notice that for each j,  $1 \le j \le n$ ,

that for each 
$$j, 1 \leq j \leq n$$
, 
$$\begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \\ 1 \\ 0 \end{bmatrix} \leftarrow jth \text{ row}$$

$$= 0\vec{c}_1 + 0\vec{c}_2 + \dots + 0\vec{c}_{j-1} + 1\vec{c}_j + 0\vec{c}_{j+1} + \dots + 0\vec{c}_n$$

$$= \vec{c}_j.$$

Thus  $\vec{c}_j \in \text{im}(A)$  for each  $j, 1 \leq j \leq n$ . It follows that

$$\operatorname{span}\{\vec{c}_1,\vec{c}_2,\ldots,\vec{c}_n\}\subseteq\operatorname{im}(A),$$

and therefore

$$\operatorname{im}(A) = \operatorname{span}\{\vec{c}_1, \vec{c}_2, \dots, \vec{c}_n\}.$$