

## Section 2.8

Subspaces of  $\mathbf{R}^n$

# Subspaces: Motivation and examples

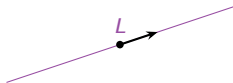
## Example

The subset  $\{0\}$ : this subspace contains only one vector.

## Example

A line  $L$  *through the origin*:

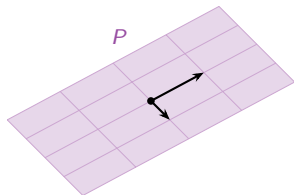
this contains the span of any vector in  $L$ .



## Example

A plane  $P$  *through the origin*:

this contains the span of any two vectors in  $P$ .



## Example

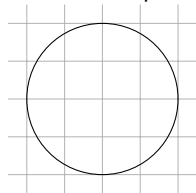
All of  $\mathbf{R}^n$ :

this contains 0, and is closed under addition and scalar multiplication.

- ▶ The **span was our first example** of subspace:  $\text{Span}\{v_1, \dots, v_p\}$ .
- ▶ But in general, subspaces are not defined by *'the generating vectors'*

## Subsets vs. Subspaces

A *subset* of  $\mathbf{R}^n$  is *any collection* of vectors whatsoever.  
All non-examples of subspaces will be still subsets of  $\mathbf{R}^n$ .



Subset: *yes*

Subspace: *no*

A **subspace** is a special kind of subset, which satisfies *three defining properties*:

1. "not empty"
2. "closed under addition"
3. "closed under  $\times$  scalars"

# Non-Examples

## Color code

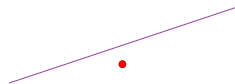
Purple: *wanna-be* 'subspaces'

Red vectors: **would have to be in** the subset too.

## Non-Example

Any set that *doesn't contain the origin*

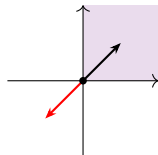
Fails condition (1).



## Non-Example

The first quadrant in  $\mathbf{R}^2$ .

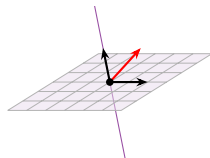
Fails close under  $\times$  scalar *only*.



## Non-Example

A line union a plane in  $\mathbf{R}^3$ .

Fails close under addition *only*.



# The Definition of Subspace

## Definition

A **subspace** of  $\mathbf{R}^n$  is a subset  $V$  of  $\mathbf{R}^n$  satisfying:

1. *The zero vector is in  $V$ .* “not empty”
2. If  $u$  and  $v$  are in  $V$ , *then  $u + v$  is also in  $V$ .* “closed under addition”
3. If  $u$  is in  $V$  and  $c$  is in  $\mathbf{R}$ , *then  $cu$  is in  $V$ .* “closed under  $\times$  scalars”

**Consequences** of definition:

- ▶ By (3), *if  $v$  is in  $V$ , then  $so is the line through  $v$ .$*
- ▶ By (2),(3), *if  $u, v$  are in  $V$ , then  $so is  $xu + yv$ , for all  $x, y \in \mathbf{R}$ .$*

A subspace  $V$  *contains the span* of any set *of vectors in  $V$ .*

# Spans are Subspaces

## Theorem

Any Span $\{v_1, v_2, \dots, v_n\}$  is a subspace.

## Check:

1.  $0 = 0v_1 + 0v_2 + \dots + 0v_n$  is in the span.
2. If, say,  $u = 3v_1 + 4v_2$  and  $v = -v_1 - 2v_2$ , then

$$u + v = 3v_1 + 4v_2 - v_1 - 2v_2 = 2v_1 + 2v_2$$

is also in the span.

3. Similarly, if  $u$  is in the span, then so is  $cu$  for any scalar  $c$ .

## Definition

If  $V = \text{Span}\{v_1, v_2, \dots, v_n\}$ , we say that  $V$  is the subspace **generated by** or **spanned by** the vectors  $v_1, v_2, \dots, v_n$ .

!!!

Every span is a subspace but also every subspace is a span.

How would you *find the generating vectors*?

## Poll

Why is the first quadrant of  $\mathbf{R}^2$  not a subspace? Which property(ies) does it fail?

$$Q = \left\{ \begin{pmatrix} a_1 \\ a_2 \end{pmatrix} : a_1 \geq 0, a_2 \geq 0 \right\}$$

1. The *zero vector is contained* in the first quadrant: ✓

2. It is *closed under addition*:  $\begin{pmatrix} a_1 \\ a_2 \end{pmatrix}, \begin{pmatrix} b_1 \\ b_2 \end{pmatrix} \in Q$  then

$$a_1 + b_1 \geq 0 \quad a_2 + b_2 \geq 0 \Rightarrow \begin{pmatrix} a_1 + b_1 \\ a_2 + b_2 \end{pmatrix} \in Q \quad \checkmark$$

3. It is **not closed under**  $\times$  scalar: let  $a_1, a_2 > 0$  and  $c < 0$  then

$$c \cdot a_1 < 0 \quad c \cdot a_2 < 0 \Rightarrow \begin{pmatrix} a_1 \\ a_2 \end{pmatrix} \in Q, \quad c \begin{pmatrix} a_1 \\ a_2 \end{pmatrix} \notin Q \quad \times$$

# Subspaces

## Verification

Let  $V = \left\{ \begin{pmatrix} a \\ b \end{pmatrix} \text{ in } \mathbf{R}^2 \mid ab = 0 \right\}$ . Let's check if  $V$  is a subspace or not.

1. Does  $V$  contain the **zero vector**?  $\begin{pmatrix} a \\ b \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \implies ab = 0$  ✓

3. Is  $V$  closed under **scalar multiplication**?

▶ Let  $\begin{pmatrix} a \\ b \end{pmatrix}$  be in  $V$ . (*a and b such that  $ab = 0$* ).  
Let  $c$  be a scalar.

▶ Is  $c \begin{pmatrix} a \\ b \end{pmatrix} = \begin{pmatrix} ca \\ cb \end{pmatrix}$  in  $V$ ?

Yes, since  $(ca)(cb) = c^2(ab) = c^2(0) = 0$  ✓

2. Is  $V$  closed under **addition**?

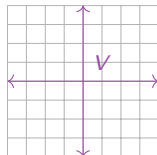
▶ Let  $\begin{pmatrix} a \\ b \end{pmatrix}$  and  $\begin{pmatrix} a' \\ b' \end{pmatrix}$  be in  $V$ . ( *$ab = 0$  and  $a'b' = 0$* ).

▶ Is  $\begin{pmatrix} a \\ b \end{pmatrix} + \begin{pmatrix} a' \\ b' \end{pmatrix} = \begin{pmatrix} a+a' \\ b+b' \end{pmatrix}$  in  $V$ ?

▶ Need to have  $(a+a')(b+b') = 0$  **always**.

However, for  $a = b' = 0$  and  $a' = b = 1$  this is not true.

▶  $\begin{pmatrix} 1 \\ 0 \end{pmatrix}$ ,  $\begin{pmatrix} 0 \\ 1 \end{pmatrix}$  are in  $V$ , but  $\begin{pmatrix} 1 \\ 1 \end{pmatrix}$  is not in  $V$ , ( $1 \cdot 1 \neq 0$ ). ✗



We conclude that  $V$  is *not* a subspace. A picture is above.



## Basis and dimension of a Subspace

Let  $V$  be a subspace of  $\mathbf{R}^n$  and  $\{v_1, v_2, \dots, v_m\}$  in  $V$  linearly independent. So that every time you 'add one' of these vectors, the *span gets bigger*.

What if  $\text{Span}\{v_1, \dots, v_m\} = V$ ?

Then any *smaller set can't span*  $V$ . If we *remove any vector*, the span gets *smaller*:

### Definition

Let  $V$  be a subspace of  $\mathbf{R}^n$ . A **basis** of  $V$  is a set of vectors  $\{v_1, v_2, \dots, v_m\}$  in  $V$  such that:

1.  $V = \text{Span}\{v_1, v_2, \dots, v_m\}$ , and
2.  $\{v_1, v_2, \dots, v_m\}$  is *linearly independent*.

The number of vectors in a basis is the **dimension** of  $V$ , and is written  $\dim V$ .

### Important

A subspace has many *different bases*, but they all have the *same number* of vectors (see the exercises in §2.9).

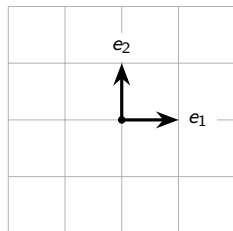
# Bases of $\mathbf{R}^2$

## Question

What is a basis for  $\mathbf{R}^2$ ?

We need two vectors that *span*  $\mathbf{R}^2$  and are *linearly independent*.  $\{e_1, e_2\}$  is one basis.

1. They span:  $\begin{pmatrix} a \\ b \end{pmatrix} = ae_1 + be_2$ .
2. They are linearly independent.

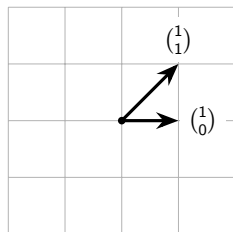


## Question

What is another basis for  $\mathbf{R}^2$ ?

Any two *nonzero vectors* that are *not collinear*.  $\left\{ \begin{pmatrix} 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 \\ 1 \end{pmatrix} \right\}$  is also a basis.

1. They span:  $\begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix}$  has a pivot in **every row**.
2. They are linearly independent:  $\begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix}$  has a pivot in **every column**.



## Bases of $\mathbf{R}^n$

The unit coordinate vectors

$$e_1 = \begin{pmatrix} 1 \\ 0 \\ \vdots \\ 0 \\ 0 \end{pmatrix}, \quad e_2 = \begin{pmatrix} 0 \\ 1 \\ \vdots \\ 0 \\ 0 \end{pmatrix}, \quad \dots, \quad e_{n-1} = \begin{pmatrix} 0 \\ 0 \\ \vdots \\ 1 \\ 0 \end{pmatrix}, \quad e_n = \begin{pmatrix} 0 \\ 0 \\ \vdots \\ 0 \\ 1 \end{pmatrix}$$

are a basis for  $\mathbf{R}^n$ . The identity matrix has columns  $e_1, e_2, \dots, e_n$ .

1. They span:  $I_n$  has a pivot in every row.
2. They are linearly independent:  $I_n$  has a pivot in every column.

In general:

$\{v_1, v_2, \dots, v_n\}$  is a basis for  $\mathbf{R}^n$  if and only if the matrix

$$A = \begin{pmatrix} | & | & \cdots & | \\ v_1 & v_2 & \cdots & v_n \\ | & | & \cdots & | \end{pmatrix}$$

has a pivot in every row and every column, i.e. if  $A$  is invertible.

# Basis of a Subspace

## Example

### Example

Let

$$V = \left\{ \begin{pmatrix} x \\ y \\ z \end{pmatrix} \text{ in } \mathbf{R}^3 \mid x + 3y + z = 0 \right\} \quad \mathcal{B} = \left\{ \begin{pmatrix} -3 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \\ -3 \end{pmatrix} \right\}.$$

Verify that  $\mathcal{B}$  is a basis for  $V$ .

0. In  $V$ : both vectors satisfy the equation, so are in  $V$

$$-3 + 3(1) + 0 = 0 \quad \text{and} \quad 0 + 3(1) + (-3) = 0.$$

1. Span: If  $\begin{pmatrix} x \\ y \\ z \end{pmatrix}$  is in  $V$ , then  $y = -\frac{1}{3}(x + z)$ , so

$$\begin{pmatrix} x \\ y \\ z \end{pmatrix} = -\frac{x}{3} \begin{pmatrix} -3 \\ 1 \\ 0 \end{pmatrix} - \frac{z}{3} \begin{pmatrix} 0 \\ 1 \\ -3 \end{pmatrix}.$$

2. Linearly independent:

$$c_1 \begin{pmatrix} -3 \\ 1 \\ 0 \end{pmatrix} + c_2 \begin{pmatrix} 0 \\ 1 \\ -3 \end{pmatrix} = \mathbf{0} \implies \begin{pmatrix} -3c_1 \\ c_1 + c_2 \\ -3c_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} \implies c_1 = c_2 = 0.$$

## Subspaces of a transformation

An  $m \times n$  matrix  $A$  naturally gives rise to *two* subspaces.

### Definition

The *column space* of  $A$  is the subspace of  $\mathbf{R}^m$  spanned by the columns of  $A$ . It is *written*  $\text{Col } A$ .

The **null space** of  $A$  is a subspace of  $\mathbf{R}^n$  containing the set of all solutions of the homogeneous equation  $Ax = 0$ :

$$\text{Nul } A = \{x \text{ in } \mathbf{R}^n \mid Ax = 0\}.$$

### Some remarks:

- ▶ The column space is the range (as opposed to the codomain) of the transformation  $T(x) = Ax$ .
- ▶ The column space is defined as a span, so we know it is a subspace.
- ▶ For the null space is easier to verify it is a subspace than find its generators. (This is one reason subspaces are so useful.)

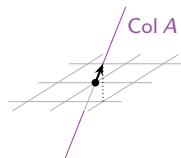
# Column Space and Null Space

Example

$$\text{Let } A = \begin{pmatrix} 1 & 1 \\ 1 & 1 \\ 1 & 1 \end{pmatrix}.$$

Let's compute the *column space*:

$$\text{Col } A = \text{Span} \left\{ \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}, \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \right\} = \text{Span} \left\{ \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \right\}.$$



This *is a line in  $\mathbb{R}^3$* .

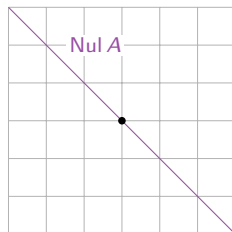
Let's compute the **null space**:

$$A \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} x+y \\ x+y \\ x+y \end{pmatrix}.$$

This **zero** if and only if  $x = -y$ . So

$$\text{Nul } A = \left\{ \begin{pmatrix} x \\ y \end{pmatrix} \text{ in } \mathbb{R}^2 \mid y = -x \right\}.$$

This defines **a line in  $\mathbb{R}^2$** :



## Verify: The null space is a subspace and a span

Check that the null space is a subspace:

1.  $0$  is in  $\text{Nul } A$  because  $A0 = 0$ .
2. If  $u$  and  $v$  are in  $\text{Nul } A$ , then  $Au = 0$  and  $Av = 0$ . Hence
$$A(u + v) = Au + Av = 0,$$
so  $u + v$  is in  $\text{Nul } A$ .
3. If  $u$  is in  $\text{Nul } A$ , then  $Au = 0$ . For any scalar  $c$ ,  $A(cu) = cAu = 0$ . So  $cu$  is in  $\text{Nul } A$ .

### Question

How to find vectors which span the null space? **Answer:** Parametric vector form!

We know that the *solution set to  $Ax = 0$*  has a parametric form that *looks like*

$$x_3 \begin{pmatrix} 1 \\ 2 \\ 1 \\ 0 \end{pmatrix} + x_4 \begin{pmatrix} -2 \\ 3 \\ 0 \\ 1 \end{pmatrix} \quad \text{if, say, } x_3 \text{ and } x_4 \text{ are the free variables. So} \quad \text{Nul } A = \text{Span} \left\{ \begin{pmatrix} 1 \\ 2 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} -2 \\ 3 \\ 0 \\ 1 \end{pmatrix} \right\}.$$

Refer back to the slides for §1.5 (Solution Sets).

# Find Null Space as a Span

## Example

Find vector(s) that span the null space of  $A = \begin{pmatrix} 1 & 1 \\ 1 & 1 \\ 1 & 1 \end{pmatrix}$ .

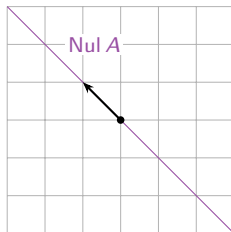
The *reduced row echelon* form is  $\begin{pmatrix} 1 & 1 \\ 0 & 0 \\ 0 & 0 \end{pmatrix}$ .

This *gives the equation*  $x + y = 0$ , or

$$\begin{array}{l} x = -y \\ y = y \end{array} \xrightarrow{\text{parametric vector form}} \begin{pmatrix} x \\ y \end{pmatrix} = y \begin{pmatrix} -1 \\ 1 \end{pmatrix}.$$

The null space is

$$\text{Nul } A = \text{Span} \left\{ \begin{pmatrix} -1 \\ 1 \end{pmatrix} \right\}.$$





# Basis for Nul $A$

## Fact

The vectors in the parametric vector form of the general solution to  $Ax = 0$  always form a basis for  $\text{Nul } A$ .

## Example

$$A = \begin{pmatrix} 1 & 2 & 0 & -1 \\ -2 & -3 & 4 & 5 \\ 2 & 4 & 0 & -2 \end{pmatrix} \xrightarrow{\text{rref}} \begin{pmatrix} 1 & 0 & -8 & -7 \\ 0 & 1 & 4 & 3 \\ 0 & 0 & 0 & 0 \end{pmatrix}$$

parametric vector form  $\xrightarrow{\text{~~~~~}}$

$$x = x_3 \begin{pmatrix} 8 \\ -4 \\ 1 \\ 0 \end{pmatrix} + x_4 \begin{pmatrix} 7 \\ -3 \\ 0 \\ 1 \end{pmatrix} \xrightarrow{\text{basis of Nul } A} \left\{ \begin{pmatrix} 8 \\ -4 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 7 \\ -3 \\ 0 \\ 1 \end{pmatrix} \right\}$$

1. The vectors span  $\text{Nul } A$  by construction (every solution to  $Ax = 0$  has this form).
2. Can you see why they are linearly independent? (Look at the last two rows.)

## Basis for Col A

Fact

The *pivot columns* of  $A$  always form a basis for Col  $A$ .

**Warning:** It is the pivot columns of the *original matrix*  $A$ , **not the row-reduced** form. (Row reduction changes the column space.)

Example

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pivot columns = basis  $\longleftrightarrow$  pivot columns in rref

So a basis for Col  $A$  is

$$\left\{ \begin{pmatrix} 1 \\ -2 \\ 2 \end{pmatrix}, \begin{pmatrix} 2 \\ -3 \\ 4 \end{pmatrix} \right\}.$$

**Why?** End of §2.8, or ask in office hours.

# Subspaces

## Summary

How do you check if a subset is a subspace?

- ▶ Is it a span?
- ▶ Is it all of  $\mathbf{R}^n$  or the zero subspace  $\{0\}$ ?  
*Can it be written as*
- ▶ a span?
- ▶ the column space of a matrix?
- ▶ the null space of a matrix?
- ▶ a type of subspace that we'll learn about later (eigenspaces, ...)?

If so, then it's automatically a subspace.

**If all else fails:**

- ▶ Can you *verify directly* that it satisfies the *three defining properties*?

## Subspaces of a transformation

Recall: a **basis** of a subspace  $V$  is a set of vectors that

- ▶ *spans*  $V$  and
- ▶ is *linearly independent*.

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

- ▶ The *column space* of  $A$  is the subspace of  $\mathbf{R}^m$  spanned by the columns of  $A$ . It is *written*  $\text{Col } A$ .
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$$\text{Nul } A = \{x \text{ in } \mathbf{R}^n \mid Ax = 0\}.$$

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parametric vector form  $\xrightarrow{\text{~~~~~}}$

$$x = x_3 \begin{pmatrix} 8 \\ -4 \\ 1 \\ 0 \end{pmatrix} + x_4 \begin{pmatrix} 7 \\ -3 \\ 0 \\ 1 \end{pmatrix} \xrightarrow{\text{basis of Nul } A} \left\{ \begin{pmatrix} 8 \\ -4 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 7 \\ -3 \\ 0 \\ 1 \end{pmatrix} \right\}$$

1. Every solution to  $Ax = 0$  has this form. So the *vectors span Nul  $A$*  by construction.
2. Look at the *last two rows of the basis*. Can you see why *they are linearly independent*?

## Basis for Col A

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The *pivot columns* of  $A$  always form a basis for Col  $A$ .

**Warning:** It is the pivot columns of the **original matrix  $A$** , *not the row-reduced* form. (Row reduction changes the column space.)

### Example

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pivot columns = basis  $\longleftrightarrow$  pivot columns in rref

So a basis for Col  $A$  is

$$\left\{ \begin{pmatrix} 1 \\ -2 \\ 2 \end{pmatrix}, \begin{pmatrix} 2 \\ -3 \\ 4 \end{pmatrix} \right\}.$$

**Why?** End of §2.8, or ask in office hours.