## Section 2.8

Subspaces of  $\mathbf{R}^n$ 

#### Motivation

Today we will discuss **subspaces** of  $\mathbf{R}^n$ .

A subspace turns out to be the same as a span, except we don't know *which* vectors it's the span of.

This arises naturally when you have, say, a plane through the origin in  $\mathbb{R}^3$  which is *not* defined (a priori) as a span, but you still want to say something about it.



### 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.
- 2. If u and v are in V, then u + v is also in V.
- 3. If u is in V and c is in  $\mathbf{R}$ , then cu is in V.

"not empty" "closed under addition"

"closed under  $\times$  scalars"

Fast-forward Every span is a subspace.

A subspace is a span of some vectors, but you haven't computed what those vectors are yet.

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# "closed under addition"

"not empty"

"closed under imes scalars"

#### What does this mean?

- If v is in V, then all scalar multiples of v are in V by (3). That is, the line through v is in V.
- If u, v are in V, then xu and yv are in V for scalars x, y by (3). So xu + yv is in V by (2). So Span{u, v} is contained in V.
- ► Likewise, if v<sub>1</sub>, v<sub>2</sub>,..., v<sub>n</sub> are all in V, then Span{v<sub>1</sub>, v<sub>2</sub>,..., v<sub>n</sub>} is contained in V: a subspace contains the span of any set of vectors in it.

If you pick enough vectors in V, eventually their span will fill up V, so:

A subspace is a span of some set of vectors in it.

#### Examples

#### 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 vectors in P.



#### Example

All of  $\mathbf{R}^n$ : this contains 0, and is closed under addition and scalar multiplication.

#### Example

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

Note these are all pictures of spans! (Line, plane, space, etc.)

A subset of  $\mathbf{R}^n$  is any collection of vectors whatsoever.

All of the following non-examples are still subsets.

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



#### Non-Examples

#### Non-Example

A line L (or any other set) that doesn't contain the origin is not a subspace. Fails: 1.

#### Non-Example

A circle C is not a subspace. Fails: 1,2,3. Think: a circle isn't a "linear space."

#### Non-Example

The first quadrant in  $\mathbf{R}^2$  is not a subspace. Fails: 3 only.

#### Non-Example

A line union a plane in  $\mathbf{R}^3$  is not a subspace. Fails: 2 only.



#### Spans are Subspaces

#### Theorem

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

Every subspace is a span, and every span is a subspace.

#### 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  $v_1, v_2, \dots, v_n$ .

#### Check:

- 1.  $0 = 0v_1 + 0v_2 + \cdots + 0v_n$  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.

#### Poll

Which of the following are subspaces?

- A. The empty set  $\{\}$ .
- B. The solution set to a homogeneous system of linear equations.
- C. The solution set to an inhomogeneous system of linear equations.
- D. The set of all vectors in  $\mathbf{R}^n$  with rational (fraction) coordinates. For the ones which are not subspaces, which property(ies) do they not satisfy?

- A. This is not a subspace: it does not contain the zero vector.
- B. This is a subspace: the solution set is a span, produced by finding the parametric vector form of the solution.
- C. This is not a subspace: it does not contain 0.
- D. This is not a subspace: it is not closed under multiplication by scalars (e.g. by  $\pi$ ).

#### Subspaces Verification

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

1. Does V contain the zero vector?  $\binom{a}{b} = \binom{0}{0} \implies ab = 0$ 

- 3. Is V closed under scalar multiplication?
  - Let <sup>a</sup><sub>b</sub> be (an unknown vector) in V.
  - This means a and b are numbers 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?
  - This means: (ca)(cb) = 0.
  - Well,  $(ca)(cb) = c^2(ab) = c^2(0) = 0$
- 2. Is V closed under addition?
  - Let  $\binom{a}{b}$  and  $\binom{a'}{b'}$  be (unknown vectors) in V.
  - This means: ab = 0 and a'b' = 0. Is  $\binom{a}{b} + \binom{a'}{b'} = \binom{a+a'}{b+b'}$  in V?

  - ► This means: (a + a')(b + b') = 0.
  - This is not true for all such a, a', b, b': for instance,  $\begin{pmatrix} 1 \\ 0 \end{pmatrix}$  and  $\begin{pmatrix} 0 \\ 1 \end{pmatrix}$  are in V, but their sum  $\binom{1}{0} + \binom{0}{1} = \binom{1}{1}$  is not in V, because  $1 \cdot 1 \neq 0$ .

We conclude that V is not a subspace. A picture is above. (It doesn't look like a span.)



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

#### Definition

- ► The column space of A is the subspace of R<sup>m</sup> spanned by the columns of A. It is written Col A.
- The **null space** of A is the set of all solutions of the homogeneous equation Ax = 0:

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

This is a subspace of  $\mathbf{R}^n$ .

The column space is defined as a span, so we know it is a subspace. It is the range (as opposed to the codomain) of the transformation T(x) = Ax.

Check that the null space is a subspace:

- 1. 0 is in Nul A because A0 = 0.
- 2. If u and v are in Nul A, then Au = 0 and Av = 0. Hence

$$A(u+v)=Au+Av=0,$$

so u + v is in Nul A.

 If u is in Nul A, then Au = 0. For any scalar c, A(cu) = cAu = 0. So cu is in Nul A.

# Column Space and Null Space Example

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

Let's compute the column space:

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

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

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

The reduced row echelon form of A is  $\begin{pmatrix} 2 \\ 0 \\ 0 \end{pmatrix}$ 

This gives the equation x + y = 0, or

$$x = -y$$
 parametric vector form  
 $y = y$ 

Hence the null space is Span $\{\binom{-1}{1}\}$ , a line in  $\mathbb{R}^2$ .

$$\begin{pmatrix} 1 & 1 \\ 0 & 0 \\ 0 & 0 \end{pmatrix}.$$
$$\begin{pmatrix} x \\ y \end{pmatrix} = y \begin{pmatrix} -1 \\ 1 \end{pmatrix}.$$





The column space of a matrix A is defined to be a span (of the columns).

The null space is defined to be the solution set to Ax = 0. It is a subspace, so it is a span.

#### 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

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

Refer back to the slides for  $\S1.5$  (Solution Sets).

Note: It is much easier to define the null space first as a subspace, then find spanning vectors *later*, if we need them. This is one reason subspaces are so useful.

#### How do you check if a subset is a subspace?

- Is it a span? Can it be written as a span?
- Can it be written as the column space of a matrix?
- Can it be written as the null space of a matrix?
- ▶ Is it all of **R**<sup>n</sup> or the zero subspace {0}?
- Can it be written as 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?

What is the *smallest number* of vectors that are needed to span a subspace?

#### Definition

Let V be a subspace of  $\mathbb{R}^n$ . 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.

Why is a basis the smallest number of vectors needed to span?

Recall: linearly independent means that every time you add another vector, the span gets bigger.

Hence, if we remove any vector, the span gets *smaller*: so any smaller set can't span V.



A subspace has *many different* bases, but they all have the same number of vectors (see the exercises in  $\S2.9$ ).

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

#### 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:  $\binom{a}{b} = ae_1 + be_2$ .
- 2. They are linearly independent because they are not collinear.

#### Question

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

Any two nonzero vectors that are not collinear.  $\left\{\binom{1}{0},\binom{1}{1}\right\}$  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

$$e_{1} = \begin{pmatrix} 1\\0\\\vdots\\0\\0 \end{pmatrix}, \quad e_{2} = \begin{pmatrix} 0\\1\\\vdots\\0\\0 \end{pmatrix}, \quad \dots, \quad e_{n-1} = \begin{pmatrix} 0\\0\\\vdots\\1\\0 \end{pmatrix}, \quad e_{n} = \begin{pmatrix} 0\\0\\\vdots\\1\\0 \end{pmatrix}$$

are a basis for  $\mathbb{R}^n$ . The identity matrix has columns  $e_1, e_2, \ldots, e_n$ . 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:  $\{v_1, v_2, \ldots, v_n\}$  is a basis for  $\mathbf{R}^n$  if and only if the matrix

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

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

Sanity check: we have shown that dim  $\mathbf{R}^n = n$ .

# Basis of a Subspace Example

#### Example

Let

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

Verify that  $\mathcal{B}$  is a basis for V. (So dim V = 2: it is a *plane*.) [interactive]

0. In V: both vectors are in V because

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

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

#### Basis for Nul A

- Fact

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

#### Example

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

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

- 1. The vectors span 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

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

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

#### Example

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

pivot columns = basis { pivot columns in rref

So a basis for Col A is

Fact

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

Why? See slides on linear independence.

#### Summary

- ► A subspace is the same as a span of some number of vectors, but we haven't computed the vectors yet.
- To any matrix is associated two subspaces, the column space and the null space:

 $\operatorname{Col} A = \operatorname{the span}$  of the columns of A Nul  $A = \operatorname{the solution}$  set of Ax = 0.

- ► A **basis** of a subspace is a minimal set of spanning vectors; the **dimension** of *V* is the number of vectors in any basis.
- ► The pivot columns form a basis for Col *A*, and the parametric vector form produces a basis for Nul *A*.

Warning These are not the official definitions!