## Warm-up

## Question

Do you understand the following sentence?
The set of $2 \times 2$ symmetric matrices is a subspace of the vector space of $2 \times 2$ matrices.

## Overview

Last time we defined an abstract vector space as a set of objects that satisfy 10 axioms. We saw that although $\mathbb{R}^{n}$ is a vector space, so is the set of polynomials of a bounded degree and the set of all $n \times n$ matrices. We also defined a subspace to be a subset of a vector space which is a vector space in its own right.

To check if a subset of a vector space is a subspace, you need to check that it contains the zero vector and is closed under addition and scalar multiplication.

Recall from 1013 that a matrix has two special subspaces associated to it: the null space and the column space.

## Question

How do the null space and column space generalise to abstract vector spaces?
(Lay, §4.2)

## Matrices and systems of equations

Recall the relationship between a matrix and a system of linear equations:
Let $A=\left[\begin{array}{lll}a_{1} & a_{2} & a_{3} \\ a_{4} & a_{5} & a_{6}\end{array}\right]$ and let $\mathbf{b}=\left[\begin{array}{l}b_{1} \\ b_{2}\end{array}\right]$.
The equation $A \mathbf{x}=\mathbf{b}$ corresponds to the system of equations

$$
\begin{aligned}
& a_{1} x+a_{2} y+a_{3} z=b_{1} \\
& a_{4} x+a_{5} y+a_{6} z=b_{2} .
\end{aligned}
$$

We can find the solutions by row-reducing the augmented matrix

$$
\left[\begin{array}{llll}
a_{1} & a_{2} & a_{3} & b_{1} \\
a_{4} & a_{5} & a_{6} & b_{2}
\end{array}\right]
$$

to reduced echelon form.

## The null space of a matrix

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

## Definition

The null space of $A$ is the set of all solutions to the homogeneous equation $A \mathbf{x}=\mathbf{0}$ :

Nul $A=\left\{\mathbf{x}: \mathbf{x} \in \mathbb{R}^{n}\right.$ and $\left.A \mathbf{x}=\mathbf{0}\right\}$.

## Example 1

Let $A=\left[\begin{array}{ccc}1 & 0 & 4 \\ 0 & 1 & -3\end{array}\right]$.
Then the null space of $A$ is the set of all scalar multiples of $v=\left[\begin{array}{c}-4 \\ 3 \\ 1\end{array}\right]$.
We can check easily that $A \mathbf{v}=\mathbf{0}$.
Furthermore, $A(t \mathbf{v})=t A \mathbf{v}=t \mathbf{0}=\mathbf{0}$, so $t \mathbf{v} \in \operatorname{Nul} A$.
To see that these are the only vectors in Nul A, solve the associated homogeneous system of equations.

## The null space theorem

Theorem (Null Space is a Subspace)
The null space of an $m \times n$ matrix $A$ is a subspace of $\mathbb{R}^{n}$.

This implies that the set of all solutions to a system of $m$ homogeneous linear equations in $n$ unknowns is a subspace of $\mathbb{R}^{n}$.

## The null space theorem

Proof Since $A$ has $n$ columns, Nul $A$ is a subset of $\mathbb{R}^{n}$. To show a subset is a subspace, recall that we must verify 3 axioms:

## The null space theorem

Proof Since $A$ has $n$ columns, Nul $A$ is a subset of $\mathbb{R}^{n}$. To show a subset is a subspace, recall that we must verify 3 axioms:

- $\mathbf{0} \in \mathrm{Nu} A$ because $A \mathbf{0}=\mathbf{0}$.
- Let $\mathbf{u}$ and $\mathbf{v}$ be any two vectors in Nul $A$. Then

$$
A \mathbf{u}=\mathbf{0} \quad \text { and } \quad A \mathbf{v}=\mathbf{0} .
$$

Therefore

$$
A(\mathbf{u}+\mathbf{v})=A \mathbf{u}+A \mathbf{v}=\mathbf{0}+\mathbf{0}=\mathbf{0} .
$$

This shows that $\mathbf{u}+\mathbf{v} \in \operatorname{Nul} A$.

- If $c$ is any scalar, then

$$
A(c \mathbf{u})=c(A \mathbf{u})=c \mathbf{0}=\mathbf{0} .
$$

This shows that $c \mathbf{u} \in \operatorname{Nul} A$.
This proves that $\operatorname{Nul} A$ is a subspace of $\mathbb{R}^{n}$.

## Example 2

Let $W=\left\{\left[\begin{array}{l}r \\ s \\ t \\ u\end{array}\right]: \begin{array}{l}3 s-4 u=5 r+t \\ 3 r+2 s-5 t=4 u\end{array}\right\}$ Show that $W$ is a subspace.

## Example 2

Let $W=\left\{\left[\begin{array}{l}r \\ s \\ t \\ u\end{array}\right]: \begin{array}{l}3 s-4 u=5 r+t \\ 3 r+2 s-5 t=4 u\end{array}\right\}$ Show that $W$ is a subspace. Hint: Find a matrix $A$ such that $\operatorname{Nul} A=W$.

## Example 2

Let $W=\left\{\left[\begin{array}{l}r \\ s \\ t \\ u\end{array}\right]: \begin{array}{l}3 s-4 u=5 r+t \\ 3 r+2 s-5 t=4 u\end{array}\right\}$ Show that $W$ is a subspace.
Hint: Find a matrix $A$ such that Nul $A=W$.
If we rearrange the equations given in the description of $W$ we get

$$
\begin{aligned}
-5 r+3 s-t-4 u & =0 \\
3 r+2 s-5 t-4 u & =0
\end{aligned}
$$

So if $A$ is the matrix $A=\left[\begin{array}{cccc}-5 & 3 & -1 & -4 \\ 3 & 2 & -5 & -4\end{array}\right]$, then $W$ is the null space of $A$, and by the Null Space is a Subspace Theorem, $W$ is a subspace of $\mathbb{R}^{4}$.

## An explicit description of Nul $A$

The span of any set of vectors is a subspace. We can always find a spanning set for Nul A by solving the associated system of equations. (See Lay §1.5).

## The column space of a matrix

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

## Definition

The column space of $A$ is the set of all linear combinations of the columns of $A$.
If $A=\left[\begin{array}{llll}\mathbf{a}_{1} & \mathbf{a}_{2} & \cdots & \mathbf{a}_{n}\end{array}\right]$, then

$$
\operatorname{Col} A=\operatorname{Span}\left\{\mathbf{a}_{1}, \mathbf{a}_{2}, \ldots, \mathbf{a}_{n}\right\} .
$$

## The column space of a matrix

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

## Definition

The column space of $A$ is the set of all linear combinations of the columns of $A$.
If $A=\left[\begin{array}{llll}\mathbf{a}_{1} & \mathbf{a}_{2} & \cdots & \mathbf{a}_{n}\end{array}\right]$, then

$$
\operatorname{Col} A=\operatorname{Span}\left\{\mathbf{a}_{1}, \mathbf{a}_{2}, \ldots, \mathbf{a}_{n}\right\} .
$$

## Theorem

The column space of an $m \times n$ matrix $A$ is a subspace of $\mathbb{R}^{m}$.
Why?

## Example 3

## Suppose

$$
W=\left\{\left[\begin{array}{c}
3 a+2 b \\
7 a-6 b \\
-8 b
\end{array}\right]: a, b \in \mathbb{R}\right\} .
$$

Find a matrix $A$ such that $W=\operatorname{Col} A$.

## Example 3

Suppose

$$
W=\left\{\left[\begin{array}{c}
3 a+2 b \\
7 a-6 b \\
-8 b
\end{array}\right]: a, b \in \mathbb{R}\right\} .
$$

Find a matrix $A$ such that $W=\operatorname{Col} A$.

$$
W=\left\{a\left[\begin{array}{l}
3 \\
7 \\
0
\end{array}\right]+b\left[\begin{array}{c}
2 \\
-6 \\
-8
\end{array}\right]: a, b \in \mathbb{R}\right\}=\operatorname{Span}\left\{\left[\begin{array}{l}
3 \\
7 \\
0
\end{array}\right],\left[\begin{array}{c}
2 \\
-6 \\
-8
\end{array}\right]\right\}
$$

Put $A=\left[\begin{array}{cc}3 & 2 \\ 7 & -6 \\ 0 & -8\end{array}\right]$. Then $W=\operatorname{Col} A$.

Another equivalent way to describe the column space is

$$
\operatorname{Col} A=\left\{A \mathbf{x}: \mathbf{x} \in \mathbb{R}^{n}\right\}
$$

## Example 4

Let

$$
\mathbf{u}=\left[\begin{array}{c}
6 \\
7 \\
1 \\
-4
\end{array}\right], \quad A=\left[\begin{array}{ccc}
5 & -5 & -9 \\
8 & 8 & -6 \\
-5 & -9 & 3 \\
3 & -2 & -7
\end{array}\right]
$$

Does $\mathbf{u}$ lie in the column space of $A$ ?
We just need to answer: does $A \mathbf{x}=\mathbf{u}$ have a solution?

Consider the following row reduction:

$$
\left[\begin{array}{ccc|c}
5 & -5 & -9 & 6 \\
8 & 8 & -6 & 7 \\
-5 & -9 & 3 & 1 \\
3 & -2 & -7 & -4
\end{array}\right] \xrightarrow{\text { rref }}\left[\begin{array}{ccc|c}
1 & 0 & 0 & 11 / 2 \\
0 & 1 & 0 & -2 \\
0 & 0 & 1 & 7 / 2 \\
0 & 0 & 0 & 0
\end{array}\right]
$$

We see that the system $A \mathbf{x}=\mathbf{u}$ is consistent.
This means that the vector $\mathbf{u}$ can be written as a linear combination of the columns of $A$.

Consider the following row reduction:

$$
\left[\begin{array}{ccc|c}
5 & -5 & -9 & 6 \\
8 & 8 & -6 & 7 \\
-5 & -9 & 3 & 1 \\
3 & -2 & -7 & -4
\end{array}\right] \xrightarrow{\text { rref }}\left[\begin{array}{ccc|c}
1 & 0 & 0 & 11 / 2 \\
0 & 1 & 0 & -2 \\
0 & 0 & 1 & 7 / 2 \\
0 & 0 & 0 & 0
\end{array}\right]
$$

We see that the system $A \mathbf{x}=\mathbf{u}$ is consistent.
This means that the vector $\mathbf{u}$ can be written as a linear combination of the columns of $A$.

Thus $\mathbf{u}$ is contained in the Span of the columns of $A$, which is the column space of $A$. So the answer is YES!

## Comparing $\operatorname{Nul} A$ and $\mathrm{Col} A$

Example 5
Let $A=\left[\begin{array}{ccccc}4 & 5 & -2 & 6 & 0 \\ 1 & 1 & 0 & 1 & 0\end{array}\right]$.

- The column space of $A$ is a subspace of $\mathbb{R}^{k}$ where $k=$ $\qquad$ .
- The null space of $A$ is a subspace of $\mathbb{R}^{k}$ where $k=$ $\qquad$ .
- Find a nonzero vector in Col $A$. (There are infinitely many.)
- Find a nonzero vector in Nul $A$.


## Comparing $\operatorname{Nul} A$ and $\operatorname{Col} A$

## Example 5

Let $A=\left[\begin{array}{ccccc}4 & 5 & -2 & 6 & 0 \\ 1 & 1 & 0 & 1 & 0\end{array}\right]$.

- The column space of $A$ is a subspace of $\mathbb{R}^{k}$ where $k=$ $\qquad$ .
- The null space of $A$ is a subspace of $\mathbb{R}^{k}$ where $k=$ $\qquad$ .
- Find a nonzero vector in Col $A$. (There are infinitely many.)
- Find a nonzero vector in Nul $A$.

For the final point, you may use the following row reduction:

$$
\left[\begin{array}{ccccc}
4 & 5 & -2 & 6 & 0 \\
1 & 1 & 0 & 1 & 0
\end{array}\right] \rightarrow\left[\begin{array}{ccccc}
1 & 1 & 0 & 1 & 0 \\
4 & 5 & -2 & 6 & 0
\end{array}\right] \rightarrow\left[\begin{array}{ccccc}
1 & 1 & 0 & 1 & 0 \\
0 & 1 & -2 & 2 & 0
\end{array}\right]
$$

Table: For any $m \times n$ matrix $A$

## Nul $A$ <br> Col $A$

1. Nul $A$ is a subspace of $\mathbb{R}^{n}$. 1.Col $A$ is a subspace of $\mathbb{R}^{m}$.
2. Any vin $\operatorname{Nul} A$ has the property that $A \mathbf{v}=\mathbf{0}$.
3. Any $\mathbf{v}$ in $\operatorname{Col} A$ has the property that the equation $A \mathbf{x}=\mathbf{v}$ is consistent.
4. Nul $A=\{\mathbf{0}\}$ if and only if 3. Col $A=\mathbb{R}^{m}$ if and only if the equation $A \mathbf{x}=\mathbf{0}$ has only the trivial solution. the equation $A \mathbf{x}=\mathbf{b}$ has a solution for every $\mathbf{b} \in \mathbb{R}^{m}$.

## Question

How does all this generalise to an abstract vector space?
An $m \times n$ matrix defines a function from $\mathbb{R}^{n}$ to $\mathbb{R}^{m}$, and the null space and column space are subspaces of the domain and range, respectively. We'd like to define the analogous notions for functions between arbitrary vector spaces.

## Linear transformations

## Definition

A linear transformation from a vector space $V$ to a vector space $W$ is a function $T: V \rightarrow W$ such that
L1. $T(\mathbf{u}+\mathbf{v})=T(\mathbf{u})+T(\mathbf{v})$ for $\mathbf{u}, \mathbf{v} \in V$;
L2. $T(c \mathbf{u})=c T(\mathbf{u})$ for $\mathbf{u} \in V, c \in \mathbb{R}$.

Matrix multiplication always defines a linear transfomation.

## Example 6

Let $A=\left[\begin{array}{ccc}1 & 0 & 2 \\ 1 & -1 & 4\end{array}\right]$. Then the mapping defined by

$$
T_{A}(\mathbf{x})=A \mathbf{x}
$$

is a linear transformation from $\mathbb{R}^{3}$ to $\mathbb{R}^{2}$.
For example

$$
T_{A}\left(\left[\begin{array}{c}
1 \\
-2 \\
3
\end{array}\right]\right)=\left[\begin{array}{ccc}
1 & 0 & 2 \\
1 & -1 & 4
\end{array}\right]\left[\begin{array}{c}
1 \\
-2 \\
3
\end{array}\right]=\left[\begin{array}{c}
7 \\
15
\end{array}\right]
$$

## Example 7

Let $T: \mathbb{P}_{2} \rightarrow \mathbb{P}_{0}$ be the map defined by

$$
T\left(a_{0}+a_{1} t+a_{2} t^{2}\right)=2 a_{0} .
$$

Then $T$ is a linear transformation.

$$
\begin{aligned}
T\left(\left(a_{0}+a_{1} t+a_{2} t^{2}\right)\right. & \left.+\left(b_{0}+b_{1} t+b_{2} t^{2}\right)\right) \\
& =T\left(\left(a_{0}+b_{0}\right)+\left(a_{1}+b_{1}\right) t+\left(a_{2}+b_{2}\right) t^{2}\right) \\
& =2\left(a_{0}+b_{0}\right) \\
& =2 a_{0}+2 b_{0} \\
& =T\left(a_{0}+a_{1} t+a_{2} t^{2}\right)+T\left(b_{0}+b_{1} t+b_{2} t^{2}\right) .
\end{aligned}
$$

$$
\begin{aligned}
T\left(c\left(a_{0}+a_{1} t+a_{2} t^{2}\right)\right) & =T\left(c a_{0}+c a_{1} t+c a_{2} t^{2}\right) \\
& =2 c a_{0} \\
& =c T\left(a_{0}+a_{1} t+a_{2} t^{2}\right)
\end{aligned}
$$

## Kernel of a linear transformation

## Definition

The kernel of a linear transformation $T: V \rightarrow W$ is the set of all vectors $\mathbf{u}$ in V such that $T(\mathbf{u})=\mathbf{0}$.
We write

$$
\operatorname{ker} T=\{\mathbf{u} \in V: T(\mathbf{u})=\mathbf{0}\}
$$

The kernel of a linear transformation $T$ is analogous to the null space of a matrix, and ker $T$ is a subspace of $V$.

If $\operatorname{ker} T=\{\mathbf{0}\}$, then $T$ is one to one.

## The range of a linear transformation

## Definition

The range of a linear transformation $T: V \rightarrow W$ is the set of all vectors in $W$ of the form $T(\mathbf{u})$ where $\mathbf{u}$ is in $V$.
We write

$$
\text { Range } T=\{\mathbf{w}: \mathbf{w}=T(\mathbf{u}) \text { for some } \mathbf{u} \in V\} .
$$

The range of a linear transformation is analogous to the columns space of a matrix, and Range $T$ is a subspace of $W$.

The linear transformation $T$ is onto if its range is all of $W$.

## Example 8

Consider the linear transformation $T: \mathbb{P}_{2} \rightarrow \mathbb{P}_{0}$ by

$$
T\left(a_{0}+a_{1} t+a_{2} t^{2}\right)=2 a_{0} .
$$

Find the kernel and range of $T$.

## Example 8

Consider the linear transformation $T: \mathbb{P}_{2} \rightarrow \mathbb{P}_{0}$ by

$$
T\left(a_{0}+a_{1} t+a_{2} t^{2}\right)=2 a_{0}
$$

Find the kernel and range of $T$.

The kernel consists of all the polynomials in $\mathbb{P}_{2}$ satisfying $2 a_{0}=0$. This is the set

$$
\left\{a_{1} t+a_{2} t^{2}\right\} .
$$

The range of $T$ is $\mathbb{P}_{0}$.

## Example 9

The differential operator $D: \mathbb{P}_{2} \rightarrow \mathbb{P}_{1}$ defined by $D(\mathbf{p}(x))=\mathbf{p}^{\prime}(x)$ is a linear transformation. Find its kernel and range.

## Example 9

The differential operator $D: \mathbb{P}_{2} \rightarrow \mathbb{P}_{1}$ defined by $D(\mathbf{p}(x))=\mathbf{p}^{\prime}(x)$ is a linear transformation. Find its kernel and range.

First we see that

$$
D\left(a+b x+c x^{2}\right)=b+2 c x
$$

So

$$
\begin{aligned}
\operatorname{ker} D & =\left\{a+b x+c x^{2}: D\left(a+b x+c x^{2}\right)=0\right\} \\
& =\left\{a+b x+c x^{2}: b+2 c x=0\right\}
\end{aligned}
$$

But $b+2 c x=0$ if and only if

## Example 9

The differential operator $D: \mathbb{P}_{2} \rightarrow \mathbb{P}_{1}$ defined by $D(\mathbf{p}(x))=\mathbf{p}^{\prime}(x)$ is a linear transformation. Find its kernel and range.

First we see that

$$
D\left(a+b x+c x^{2}\right)=b+2 c x
$$

So

$$
\begin{aligned}
\operatorname{ker} D & =\left\{a+b x+c x^{2}: D\left(a+b x+c x^{2}\right)=0\right\} \\
& =\left\{a+b x+c x^{2}: b+2 c x=0\right\}
\end{aligned}
$$

But $b+2 c x=0$ if and only if $b=2 c=0$, which implies $b=c=0$. Therefore

$$
\begin{aligned}
\operatorname{ker} D & =\left\{a+b x+c x^{2}: b=c=0\right\} \\
& =\{a: a \in \mathbb{R}\}
\end{aligned}
$$

The range of $D$ is all of $\mathbb{P}_{1}$ since every polynomial in $\mathbb{P}_{1}$ is the image under $D$ (i.e the derivative) of some polynomial in $\mathbb{P}_{2}$.
To be more specific, if $a+b x$ is in $\mathbb{P}_{1}$, then

$$
a+b x=D\left(a x+\frac{b}{2} x^{2}\right)
$$

## Example 10

Define $S: \mathbb{P}_{2} \rightarrow \mathbb{R}^{2}$ by

$$
S(\mathbf{p})=\left[\begin{array}{l}
\mathbf{p}(0) \\
\mathbf{p}(1)
\end{array}\right] .
$$

That is, if $\mathbf{p}(x)=a+b x+c x^{2}$, we have

$$
S(\mathbf{p})=\left[\begin{array}{c}
a \\
a+b+c
\end{array}\right] .
$$

Show that $S$ is a linear transformation and find its kernel and range.

Leaving the first part as an exercise to try on your own, we'll find the kernel and range of $S$.

- From what we have above, $\mathbf{p}$ is in the kernel of $S$ if and only if

$$
S(\mathbf{p})=\left[\begin{array}{c}
a \\
a+b+c
\end{array}\right]=\left[\begin{array}{l}
0 \\
0
\end{array}\right]
$$

For this to occur we must have $a=0$ and $c=-b$.
So $\mathbf{p}$ is in the kernel of $S$ if

$$
\mathbf{p}(x)=b x-b x^{2}=b\left(x-x^{2}\right)
$$

This gives ker $S=$ Span $\left\{x-x^{2}\right\}$.

- The range of $S$.

Since $S(\mathbf{p})=\left[\begin{array}{c}a \\ a+b+c\end{array}\right]$ and $a, b$ and $c$ are any real numbers, the range of $S$ is all of $\mathbb{R}^{2}$.

## Example 11

let $F: M_{2 \times 2} \rightarrow M_{2 \times 2}$ be the linear transformation defined by taking the transpose of the matrix:

$$
F(A)=A^{T} .
$$

We find the kernel and range of $F$.

## Example 11

let $F: M_{2 \times 2} \rightarrow M_{2 \times 2}$ be the linear transformation defined by taking the transpose of the matrix:

$$
F(A)=A^{T}
$$

We find the kernel and range of $F$.
We see that

$$
\begin{aligned}
\operatorname{ker} F & =\left\{A \text { in } M_{2 \times 2}: F(A)=0\right\} \\
& =\left\{A \text { in } M_{2 \times 2}: A^{T}=0\right\}
\end{aligned}
$$

But if $A^{T}=0$, then $A=\left(A^{T}\right)^{T}=0^{T}=0$. It follows that ker $F=0$. For any matrix $A$ in $M_{2 \times 2}$, we have $A=\left(A^{T}\right)^{T}=F\left(A^{T}\right)$. Since $A^{T}$ is in $M_{2 \times 2}$ we deduce that Range $F=M_{2 \times 2}$.

## Example 12

Let $S: \mathbb{P}_{1} \rightarrow \mathbb{R}$ be the linear transformation defined by

$$
S(\mathbf{p}(x))=\int_{0}^{1} \mathbf{p}(x) d x .
$$

We find the kernel and range of $S$.
In detail, we have

$$
\begin{aligned}
S(a+b x) & =\int_{0}^{1}(a+b x) d x \\
& =\left[a x+\frac{b}{2} x^{2}\right]_{0}^{1} \\
& =a+\frac{b}{2}
\end{aligned}
$$

Therefore,

$$
\begin{aligned}
\operatorname{ker} S & =\{a+b x: S(a+b x)=0\} \\
& =\left\{a+b x: a+\frac{b}{2}=0\right\} \\
& =\left\{a+b x: a=-\frac{b}{2}\right\} \\
& =\left\{-\frac{b}{2}+b x\right\}
\end{aligned}
$$

Geometrically, ker $S$ consists of all those linear polynomials whose graphs have the property that the area between the line and the $x$-axis is equally distributed above and below the axis on the interval $[0,1]$.

The range of $S$ is $\mathbb{R}$, since every number can be obtained as the image under $S$ of some polynomial in $\mathbb{P}_{1}$. For example, if $a$ is an arbitrary real number, then

$$
\int_{0}^{1} a d x=[a x]_{0}^{1}=a-0=a
$$

