

## Lecture 14b

### Rank and Dimension (No row reduce)

# Review

So far, we have two main algorithms for finding bases and the dimension of special subspaces.

## Finding a basis and dimension for standard subspaces

Subspace	Method to find one basis	Dimension
1. Image of A	Columns with L1 in REF	rank
2. Span of $\{v_1, \dots, v_n\}$	$= \text{im}(\text{concatenation})$ , use $\uparrow$	$\uparrow$
3. Kernel of A	Vectors in general solution	width $-$ rank
4. Solutions to HSLE	$= \text{ker}(\text{coeff. matrix})$ , use $\uparrow$	$\uparrow$
5. $\lambda$ -eigenspace of A	$= \text{ker}(A - \lambda \text{Id})$ , use $\uparrow$	$\uparrow$

In each case, the dimension is easy if we know a certain rank.

## Goal

Compute rank and dimension without row reduction

For a fixed matrix  $A$ , we have two simple formulas.

$$\dim(\text{im}(A)) = \text{rank}(A)$$

$$\dim(\text{ker}(A)) = \text{width}(A) - \text{rank}(A) \quad +$$

Each formula requires the rank of  $A$ ...but their **sum** does not.

## The Rank-Nullity Theorem

Let  $A$  be any matrix. Then

$$\underbrace{\dim(\text{im}(A))}_{\text{rank}(A)} + \underbrace{\dim(\text{ker}(A))}_{\text{'nullity' of } A} = \text{width}(A)$$

**Nullity** is an archaic word for the dimension of the kernel.

## Exercise 5

The image of the following matrix is a plane in  $\mathbb{R}^3$ .

$$A := \begin{bmatrix} 1 & -1 & 4 \\ 2 & 0 & 6 \\ 1 & 1 & 2 \end{bmatrix}$$

Find the dimension of the kernel of A.

$\text{im}(A)$  is a plane, so  $\dim(\text{im}(A)) = 2$ .

Recall that  $\dim(\text{im}(A))$  could be found by row-reducing and counting the number of columns with leading 1, so (in general)  $\dim(\text{im}(A)) = \text{rank}(A)$ . So  $\text{rank}(A) = 2$

$$\begin{aligned} \text{We've seen (in general)} \quad \dim(\ker(A)) &= \text{width}(A) - \text{rank}(A) \\ &= 3 - 2 \\ &= 1 \end{aligned}$$

Alternatively: Use the Rank-Nullity Thm  
 $\dim(\ker(A)) + \dim(\text{im}(A)) = \text{width}(A)$   
So  $\dim(\ker(A)) = \text{width}(A) - \dim(\text{im}(A))$   
 $= 3 - 2 = 1$ .

Side note: Geometrically, this means that the set of vectors sent to  $\begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$  by  $T_A$  is a line through the origin in  $\mathbb{R}^3$  space.

# Subspaces with dimension 0

What dimensions are possible, and what do they tell us?  
Let's consider several special cases.

**Definition:** The zero subspace

The **zero subspace** of  $\mathbb{R}^n$  only contains the zero vector.

Note: this is the only subspace with finitely many elements.

- If there are more than just one element in a subspace  $W$ , there must be infinitely many elements in  $W$ .

**Fact/Definition** (The subspace of dimension 0)

The only 0 dimensional subspace of  $\mathbb{R}^n$  is the **zero subspace**.

**Fact** (Matrices of rank 0)

The only  $m \times n$  matrix of rank 0 is the **zero matrix**.

- If an REF matrix has no leading 1, it's the zero matrix
- A nonzero matrix cannot be turned into the zero matrix using elementary row operations.

# Subspaces with maximum dimension

We can also consider the case of maximum dimension.

Recall Fact: A bound on dimension of subspaces

A subspace of  $\mathbb{R}^n$  has dimension at most  $n$ .

(A basis for a subspace of  $\mathbb{R}^n$  has  $n$  or fewer vectors)

Fact (The subspace of maximum dimension)

The only  $n$  dimensional subspace of  $\mathbb{R}^n$  is all of  $\mathbb{R}^n$ .

For example, if a subspace  $W$  of  $\mathbb{R}^5$  has dimension 5, then  $W$  must be the entire  $\mathbb{R}^5$  ( $W = \mathbb{R}^5$ ).

Matrices of maximal rank aren't unique, e.g. any **invertible** matrix has largest possible rank.

# Subspaces with dimension 1

## Fact (Subspaces of dimension 1)

A subspace is 1 dimensional if it consists of multiples of a non-zero vector  
(Geometrically, a line through the origin in  $n$  dimensional space)

## Fact (Matrices of rank 1)

A non-zero matrix has rank 1 if and only if all the columns are multiples of each other.

(Equivalently, a non-zero matrix has rank 1 iff all the rows are multiples of each other)

## Example

three times  $R_1$   $\rightarrow$

$\text{rank} \begin{bmatrix} 1 & 2 & -1 \\ 3 & 6 & -3 \\ -1 & -2 & -1 \end{bmatrix} = 1$

$-1$  times  $R_1$   $\rightarrow$

If we were to do row reduction, we would get  $\begin{bmatrix} 1 & 2 & -1 \\ 3 & 6 & -3 \\ -1 & -2 & -1 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 2 & -1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$

## Exercise 6

$$\text{Let } A := \begin{bmatrix} 1 & 2 \\ 2 & 1 \end{bmatrix}$$

- (a) Find the rank of the following matrix. (b) What is  $\dim(\text{im}(A))$ ?  
(c) What is the dimension of the kernel of  $A$ ?

[Let's practice applying the methods from this lecture — no row reduce]

(a) Possible rank for a  $2 \times 2$  matrix:

~~0~~  
A is not  
the zero  
matrix

~~1~~  
The rows of  $A$   
are not  
multiples of  
each other

2

$$\text{rank}(A) = 2$$

$$(b) \dim(\text{im}(A)) = \text{rank}(A) = 2$$

# of columns of REF  
with leading 1

$$(c) \dim(\text{ker}(A)) = \text{width}(A) - \text{rank}(A) \\ = 2 - 2 = 0$$

Think/remember:  
# of columns of REF  
with no leading 1

Note: This means  $\text{ker}(A)$  is the zero subspace,  $\text{ker}(A) = \left\{ \begin{bmatrix} 0 \\ 0 \end{bmatrix} \right\}$ .  
So the only vector sent to  $\begin{bmatrix} 0 \\ 0 \end{bmatrix}$  by  $T_A$  is  $\begin{bmatrix} 0 \\ 0 \end{bmatrix}$ .

## Exercise 7

$$\text{Let } B := \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 2 \\ 1 & 1 & 3 \end{bmatrix}$$

- Find the rank of  $B$ .
- What is the dimension of the image of  $B$ ?
- What is the dimension of the kernel of  $B$ ?

(a) Possible ranks:

~~0~~  
 $B$  is not  
the zero  
matrix

~~1~~  
The 1st and 2nd  
columns are  
multiple of  
each other,  
but the 3rd column  
is not a multiple  
of the 1st col.

2

~~3~~  
The 1st and 2nd columns are  
equal. Recall this  
means  $\det(B) = 0$ .  
So  $B$  is not invertible.  
So  $\text{rank}(B)$  is not maximum.

$$\text{So } \boxed{\text{rank}(B) = 2}$$

$$(b) \dim(\text{Im}(B)) = \text{rank}(B) = \boxed{2}$$

$$(c) \dim(\text{Ker}(B)) = \text{width}(B) - \text{rank}(B) \\ = 3 - 2 = \boxed{1}$$

This means  $\text{ker}(B)$  is a line through the origin in 3D space.

We can also relate dimension to containment between subspaces.

### Theorem 3: Subspaces contained in other subspaces

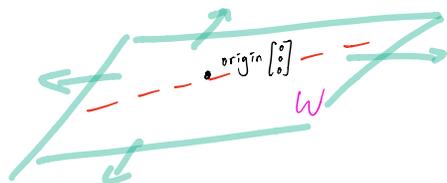
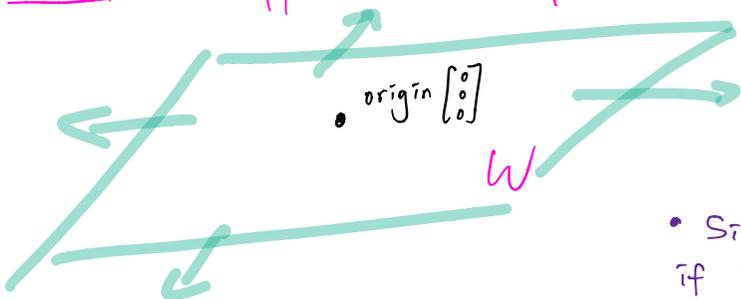
Let  $V$  and  $W$  be subspaces of  $\mathbb{R}^n$ .

(a) • If  $V$  is contained in  $W$ , then  $\dim(V) \leq \dim(W)$ .

(b) • If  $V$  is contained in  $W$  and  $\dim(V) = \dim(W)$ , then  $V = W$ .

Think of dimension as a measure of how "big" a subspace is.

Example Suppose  $W$  is a plane through the origin in  $\mathbb{R}^3$



• If  $V$  is a subspace contained in  $W$  with dimension 2, then  $V$  must be the entire plane  $W$ . (by (b))

• Since  $W$  has dimension 2, if  $V$  is contained in  $W$ , then  $V$  can have dimension 0, 1, or 2 (by (a)).

- If  $V$  has dimension 1, then  $V$  must be some line in  $W$  through the origin.

- If  $V$  has dimension 0, then  $V = \left\{ \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \right\}$ , the zero subspace of  $\mathbb{R}^3$ .

## Exercise 8

Let  $A$  and  $B$  be two matrices such that the product  $AB$  is defined.

① Show that  $\text{im}(AB)$  is contained in  $\text{im}(A)$ .

② Show that  $\text{rank}(AB) \leq \text{rank}(A)$ .

③ Show that  $\text{rank}(AB) \leq \text{rank}(B)$ . **Hint: Is there a trick to reverse the order of multiplication without changing the rank?** *I'll use transpose, since  $\text{rank}(M) = \text{rank}(M^T)$  and  $(CD)^T = D^T C^T$*

This shows the following general principle, which is virtually impossible to show from the leading 1s definition.

## Rank and matrix multiplication

$$\text{rank}(AB) \leq \min(\text{rank}(A), \text{rank}(B))$$

### Exercise 8

Let A and B be two matrices such that the product AB is defined.

- 1 Show that  $\text{im}(AB)$  is contained in  $\text{im}(A)$ .

We want to show  $\text{im}(AB) \subset \text{im}(A)$

[" $\subset$ " means: "contained in" or "is a subset of"]

Let  $v$  be in  $\text{im}(AB)$ ; that is,  
there is some  $w$  such that

$$\begin{aligned} v &= (AB)w \\ &= A(\underbrace{Bw}_{\text{(a vector)}}) \end{aligned}$$

Therefore,  $v$  is in  $\text{im}(A)$ .

Since this computation works for all  $v$  in  $\text{im}(AB)$ ,  
we can say that  $\text{im}(AB)$  is in  $\text{im}(A)$ . — the end of ① —

### Exercise 8

Let A and B be two matrices such that the product AB is defined.

- 1 Show that  $\text{im}(AB)$  is contained in  $\text{im}(A)$ .
- 2 Show that  $\text{rank}(AB) \leq \text{rank}(A)$ . ] will show (2)

Since  $\text{im}(AB)$  is contained in  $\text{im}(A)$  by part (1),  
Theorem 3 tells us that

$$\begin{aligned} \dim(\text{im}(AB)) &\leq \dim(\text{im}(A)), \\ \text{rank}(AB) &\leq \text{rank}(A). \end{aligned}$$

— the end of (2) —

### Exercise 8

Let A and B be two matrices such that the product AB is defined.

- 1 Show that  $\text{im}(AB)$  is contained in  $\text{im}(A)$ .
- 2 Show that  $\text{rank}(AB) \leq \text{rank}(A)$ .
- 3 Show that  $\text{rank}(AB) \leq \text{rank}(B)$ . **Hint: Is there a trick to reverse the order of multiplication without changing the rank?** I'll use transpose, since  $\text{rank}(M) = \text{rank}(M^T)$  and  $(AB)^T = B^T A^T$

Recall:

a) rank of a matrix equals  
rank of its transpose

$$\text{rank}(M) = \text{rank}(M^T)$$

b)  $(CD)^T = D^T C^T$ .

rank(AB)  $\stackrel{(a)}{=} \text{rank}((AB)^T)$

$\stackrel{(b)}{=} \text{rank}(B^T A^T)$

$\leq \text{rank}(B^T)$  by part (2)

$\stackrel{(a)}{=} \text{rank}(B)$

So  $\text{rank}(AB) \leq \text{rank}(B)$ .