

# ENGI 7825: Linear Algebra Review

## Subspaces, Basis, Dimension, and Rank

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## Definition: Vector Space

A **vector space** is a set  $\mathbb{V}$  which is closed under the following operations (where  $\vec{x}, \vec{y}, \vec{z} \in \mathbb{V}$  and  $c, t \in \mathbb{R}$ ):

- **vector addition:**  $\vec{x} + \vec{y} \in \mathbb{V}$
- **scalar multiplication:**  $c\vec{x} \in \mathbb{V}$

The following properties hold for any vector space:

- 1  $\vec{x} + \vec{y} \in \mathbb{V}$
- 2  $\vec{x} + \vec{y} = \vec{y} + \vec{x}$
- 3  $(\vec{x} + \vec{y}) + \vec{z} = \vec{x} + (\vec{y} + \vec{z})$
- 4 There is an element  $\vec{0} \in \mathbb{V}$  so that  $\vec{x} + \vec{0} = \vec{x} = \vec{0} + \vec{x}$
- 5 For each  $\vec{x} \in \mathbb{V}$  there is a  $-\vec{x} \in \mathbb{V}$  so that  $\vec{x} + (-\vec{x}) = \vec{0}$
- 6  $c\vec{x} \in \mathbb{V}$
- 7  $c(\vec{x} + \vec{y}) = c\vec{x} + c\vec{y}$
- 8  $(c + t)\vec{x} = c\vec{x} + t\vec{x}$
- 9  $c(t\vec{x}) = (ct)\vec{x}$
- 10  $1\vec{x} = \vec{x}$

Interestingly, the concept of vector spaces applies to non-vectors. Anything that has definitions for addition and scalar multiplication that satisfy the closure property of vector spaces qualifies.

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**Example:** Let  $P_n$  be all polynomials of degree at most  $n$ . (These can be viewed as functions  $\mathbb{R} \rightarrow \mathbb{R}$ .)

**Specific Example:**  $P_5$  is every function we can write as

$$f(t) = a_5t^5 + a_4t^4 + a_3t^3 + a_2t^2 + a_1t^1 + a_0$$

Can **add** any two of these or **multiply** by a scalar and you still get a member of  $P_5$ .

Example: if

$$f(t) = 3t^5 - 2t^4 + t^3 + 4t^2 + t^1 - 3$$

then

$$5f(t) = 15t^5 - 10t^4 + 5t^3 + 20t^2 + 5t^1 - 15$$

## Definition: Subspace

Suppose that  $\mathbb{V}$  is a vector space and  $\mathbb{U} \subset \mathbb{V}$ . That is,  $\mathbb{U}$  is contained in  $\mathbb{V}$ . Suppose further

- $\vec{0} \in \mathbb{U}$ ,
- for all  $\vec{x}, \vec{y} \in \mathbb{U}$ , the sum  $\vec{x} + \vec{y} \in \mathbb{U}$
- for all  $c \in \mathbb{R}$  and  $\vec{x} \in \mathbb{U}$ , the scalar product  $c\vec{x} \in \mathbb{U}$

Then  $\mathbb{U}$  is a **subspace** of  $\mathbb{V}$

## Theorem

*Any subspace of a vector space is itself a vector space.*

**Example:** Let  $P_5$  be the vector space of all 5<sup>th</sup> degree polynomials.  $P_4$  is a subspace of  $P_5$ .

**non-Example** Let  $\mathbb{U} \subset \mathbb{R}^2$  be the set of all  $\begin{bmatrix} x \\ y \end{bmatrix}$  so that  $x \geq 0, y \geq 0$ . Check if it's a subspace of  $\mathbb{R}^2$ :

- $\vec{0} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \in \mathbb{U} \quad \checkmark$

- Suppose  $\begin{bmatrix} x \\ y \end{bmatrix} \in \mathbb{U}$  and  $\begin{bmatrix} x' \\ y' \end{bmatrix} \in \mathbb{U}$ . Then

$$x, x' \geq 0 \implies x + x' \geq 0 \text{ and } y, y' \geq 0 \implies y + y' \geq 0.$$

Hence  $\begin{bmatrix} x \\ y \end{bmatrix} + \begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} x + x' \\ y + y' \end{bmatrix} \in \mathbb{U} \quad \checkmark$

- But, sadly, although  $\begin{bmatrix} 1 \\ 1 \end{bmatrix} \in \mathbb{U}$ , the scalar product

$$-1 \cdot \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \begin{bmatrix} -1 \\ -1 \end{bmatrix} \notin \mathbb{U}. \quad \text{☹}$$

Since we have shown some example **does not** satisfy one of the criteria,  $\mathbb{U}$  is **not** a subspace.

Suppose  $\mathbb{V}$  is any vector space and  $\vec{x}_1, \vec{x}_2, \dots, \vec{x}_k \in \mathbb{V}$ .

**Definition:** Span (we have seen this before)

$\text{Span}\{\vec{x}_1, \vec{x}_2, \dots, \vec{x}_k\}$  is the set of all elements of  $\mathbb{V}$  that can be written as a **linear combination** of  $\{\vec{x}_1, \vec{x}_2, \dots, \vec{x}_k\}$ .

**Example:** In the vector space  $\mathbb{R}^3$   $\text{Span}\{[1 \ 0 \ 0]^T, [0 \ 1 \ 0]^T\}$  is the set of all vectors in the  $xy$ -plane, which is a subspace of  $\mathbb{R}^3$ .

**Theorem**

$\text{Span}\{\vec{x}_1, \vec{x}_2, \dots, \vec{x}_k\}$  is always a subspace of  $\mathbb{V}$

We can view the multiplication of a  $m \times n$  matrix  $A$  by a  $n \times 1$  vector  $\vec{x}$  in two distinct ways.

**Dot product with rows:** Each row of  $\vec{y} = A\vec{x}$  consists of the dot product of the corresponding row of  $A$  with  $\vec{x}$ . Let the  $m$  rows of  $A$  be the  $\vec{r}$  vectors below.

$$A = \begin{bmatrix} \vec{r}_1 \\ \vec{r}_2 \\ \dots \\ \vec{r}_m \end{bmatrix}$$

$$\vec{y} = \begin{bmatrix} \vec{r}_1\vec{x} \\ \vec{r}_2\vec{x} \\ \dots \\ \vec{r}_m\vec{x} \end{bmatrix}$$

**Combination of the columns:** The resultant vector  $\vec{y} = A\vec{x}$  consists of the combination of columns of  $A$  as given by the elements of  $\vec{x}$ . Let the  $n$  columns of  $A$  be the  $\vec{c}$  vectors below.

$$A = [\vec{c}_1 \quad \vec{c}_2 \quad \dots \quad \vec{c}_n]$$
$$\vec{y} = \vec{c}_1 x_1 + \vec{c}_2 x_2 + \dots + \vec{c}_n x_n$$



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

Consider all  $\vec{u} \in \mathbb{R}^n$  so that  $A\vec{u} = \vec{0}$ . Call it the **null space of  $A$**  or **Nul  $A$** .

We now show that **Nul  $A$**  is a **subspace** of  $\mathbb{R}^n$ .

Need to check three things:

- $\vec{0} \in \text{Nul } A$  (since  $A\vec{0} = \vec{0}$ ).
- For any  $\vec{x}$  &  $\vec{y} \in \text{Nul } A$ ,

$$A(\vec{x} + \vec{y}) = A\vec{x} + A\vec{y} = \vec{0} + \vec{0} = \vec{0}.$$

Hence  $\vec{x} + \vec{y} \in \text{Nul } A$ .

- Similarly, for any  $c \in \mathbb{R}$  and  $\vec{x} \in \text{Nul } A$ ,

$$A(c\vec{x}) = cA\vec{x} = c\vec{0} = \vec{0}.$$

Hence  $c\vec{x} \in \text{Nul } A$ .

**Easy** question: Given  $\vec{u} \in \mathbb{R}^n$ , is  $\vec{u} \in \text{Nul } A$ ?

**Harder** question: Describe all of  $\text{Nul } A$ .

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Example: Let

$$A = \begin{bmatrix} 1 & 2 & 3 & 1 \\ 1 & 3 & 5 & -2 \\ 3 & 8 & 13 & -3 \end{bmatrix} \quad \vec{u} = \begin{bmatrix} 1 \\ -2 \\ 1 \\ 0 \end{bmatrix}$$

Is  $\vec{u}$  in  $\text{Nul } A$ ?

**Answer:** Just calculate  $A\vec{u}$ :  $A\vec{u} = \vec{0} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \iff \vec{u} \in \text{Nul } A$ .

Here

$$A\vec{u} = \begin{bmatrix} 1 & 2 & 3 & 1 \\ 1 & 3 & 5 & -2 \\ 3 & 8 & 13 & -3 \end{bmatrix} \begin{bmatrix} 1 \\ -2 \\ 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 1 \cdot 1 - 2 \cdot 2 + 3 \cdot 1 + 1 \cdot 0 \\ 1 \cdot 1 - 2 \cdot 3 + 5 \cdot 1 - 2 \cdot 0 \\ 1 \cdot 3 - 2 \cdot 8 + 13 \cdot 1 - 3 \cdot 0 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

Hence  $\vec{u} \in \text{Nul } A$ .

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**Harder** question: Describe **all** of  $\text{Nul } A$ . Interpretation: Find  $\vec{u}_1, \dots, \vec{u}_k \in \text{Nul } A$  so that  $\text{Span}\{\vec{u}_1, \dots, \vec{u}_k\} = \text{Nul } A$

Fancy description of an old problem we know how to do:

Solve the homogeneous system of linear equations given by  $A\vec{x} = \vec{0}$ .

**Step 1:** Convert  $A$  to reduced echelon form:

$$\begin{bmatrix} 1 & 2 & 3 & 1 \\ 1 & 3 & 5 & -2 \\ 3 & 8 & 13 & -3 \end{bmatrix} \xrightarrow{1^{\text{st}} \text{ col.}} \begin{bmatrix} 1 & 2 & 3 & 1 \\ 0 & 1 & 2 & -3 \\ 0 & 2 & 4 & -6 \end{bmatrix} \xrightarrow{2^{\text{nd}} \text{ col.}} \begin{bmatrix} 1 & 0 & -1 & 7 \\ 0 & 1 & 2 & -3 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

**Step 2:** Identify the free variables. Here they are  $x_3, x_4$ .

**Step 3:** Convert to vector equation:

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} x_3 - 7x_4 \\ -2x_3 + 3x_4 \\ x_3 \\ x_4 \end{bmatrix} = x_3 \begin{bmatrix} 1 \\ -2 \\ 1 \\ 0 \end{bmatrix} + x_4 \begin{bmatrix} -7 \\ 3 \\ 0 \\ 1 \end{bmatrix}$$

$$\text{Then Nul } A = \text{Span} \left\{ \begin{bmatrix} 1 \\ -2 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} -7 \\ 3 \\ 0 \\ 1 \end{bmatrix} \right\}$$

This spanning set is **efficient**: no proper subset spans Nul  $A$ .

As before, let  $A$  be any  $m \times n$  matrix:

Consider **all** linear combinations of the column vectors of  $A$ . It's a subset of  $\mathbb{R}^m$ , denoted **Col  $A$** .

In other words, for

$$A = [\vec{a}_1 \quad \vec{a}_2 \quad \dots \quad \vec{a}_n]$$

Col  $A$  is  $\text{Span}\{\vec{a}_1, \vec{a}_2, \dots, \vec{a}_n\}$ .

**Connection to  $\vec{y} = A\vec{x}$ :** This system is solvable if and only if  $\vec{y}$  is in the column space of  $A$

$$\vec{y} \in \text{Col } A \iff \vec{y} = x_1\vec{a}_1 + x_2\vec{a}_2 + \dots + x_n\vec{a}_n \iff \vec{y} = A \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}$$

## Describe Col A:

Answer: We've just done it – it's  $\text{Span}\{\vec{a}_1, \dots, \vec{a}_n\}$ , the span of the column vectors of  $A$ .

## Example:

$$A = \begin{bmatrix} 1 & 2 & 3 & 1 \\ 1 & 3 & 5 & -2 \\ 3 & 8 & 13 & -3 \end{bmatrix} \implies \text{Col } A = \text{Span} \left\{ \begin{bmatrix} 1 \\ 1 \\ 3 \end{bmatrix}, \begin{bmatrix} 2 \\ 3 \\ 8 \end{bmatrix}, \begin{bmatrix} 3 \\ 5 \\ 13 \end{bmatrix}, \begin{bmatrix} 1 \\ -2 \\ -3 \end{bmatrix} \right\}$$

But this is not necessarily an **efficient** description: a subset may span Col  $A$ . In this case, the first two vectors suffice:

$$\text{Col } A = \text{Span} \left\{ \begin{bmatrix} 1 \\ 1 \\ 3 \end{bmatrix}, \begin{bmatrix} 2 \\ 3 \\ 8 \end{bmatrix} \right\}$$

**Harder** question: Given  $\vec{u} \in \mathbb{R}^m$ , is  $\vec{u} \in \text{Col } A$ .

Translation: Are there  $\{x_1, \dots, x_n\}$  so that

$$\vec{u} = x_1 \vec{a}_1 + \dots + x_n \vec{a}_n?$$

Equivalently:

Is there a vector  $\vec{x} \in \mathbb{R}^n$  so that  $A\vec{x} = \vec{u}$ ?

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**Solution:** Reduce **augmented** matrix  $[\vec{a}_1 \ \vec{a}_2 \ \dots \ \vec{a}_n \ \vec{u}]$  to echelon form and see if the equations are consistent:

Example: Let

$$A = \begin{bmatrix} 1 & 2 & 3 & 1 \\ 1 & 3 & 5 & -2 \\ 3 & 8 & 13 & -3 \end{bmatrix} \quad \vec{u} = \begin{bmatrix} 3 \\ 4 \\ 11 \end{bmatrix}$$

Is  $\vec{u} \in \text{Col } A$ ?

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$$\left[ \begin{array}{cccc|c} 1 & 2 & 3 & 1 & 3 \\ 1 & 3 & 5 & -2 & 4 \\ 3 & 8 & 13 & -3 & 11 \end{array} \right] \rightarrow \left[ \begin{array}{cccc|c} 1 & 2 & 3 & 1 & 3 \\ 0 & 1 & 2 & -3 & 1 \\ 0 & 2 & 4 & -6 & 2 \end{array} \right] \rightarrow \left[ \begin{array}{cccc|c} 1 & 2 & 3 & 1 & 3 \\ 0 & 1 & 2 & -3 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{array} \right]$$

The last column says that  $0x_1 + 0x_2 + 0x_3 + 0x_4 = 0u_3$  which is consistent. However, if the bottom-right entry had been non-zero then we would have had an inconsistent equation. So in this case  $\vec{u} \in \text{Col } A$ .

Additional payoff: First two columns are the pivot columns  $\implies$  first two **of the original columns** span  $\text{Col } A$ .



## Definition

A subset  $\{\vec{v}_1, \dots, \vec{v}_k\} \subset \mathbb{V}$  is a **basis** for  $V$  if and only if it is linearly independent **and** it spans  $V$ .

**Classic example:** The set  $\{\vec{e}_1, \dots, \vec{e}_n\} \subset \mathbb{R}^n$  is a basis for the vector space  $\mathbb{R}^n$ . Where

$$\vec{e}_i = \begin{bmatrix} 0 \\ \vdots \\ 0 \\ 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

where the 1 appears in the  $i^{\text{th}}$  position. This set is called the standard basis for  $\mathbb{R}^n$ .

## Theorem

*Any finite spanning set of vectors contains a basis.*

Suppose  $\{\vec{v}_1, \dots, \vec{v}_n\} \subset \mathbb{V}$  spans  $\mathbb{V}$

- if  $\{\vec{v}_1, \dots, \vec{v}_n\}$  is linearly **independent**, then it's a basis.
- if it's linearly **dependent**, some proper subset spans.

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Two important properties of a basis:

- No proper subset of a basis is a basis (it will no longer span).
- Adding an additional vector to a basis will no longer constitute a basis (no longer linearly independent).

A basis efficiently captures most information about a vector space.

**Example:**

Consider the set of 2-vectors

$$S = \left\{ \begin{bmatrix} 1 \\ 2 \end{bmatrix}, \begin{bmatrix} 2 \\ 1 \end{bmatrix}, \begin{bmatrix} 3 \\ 3 \end{bmatrix} \right\}$$

Is this set a basis for  $\mathbb{R}^2$ ?

No. This set is not independent. But any pair of vectors from  $S$  would form a basis for  $\mathbb{R}^2$ .

The **dimension** of a subspace is the number of vectors needed to form a basis.

Its clear that the dimension of  $\mathbb{R}^n$  is  $n$ . Any vector in  $\mathbb{R}^n$  can be written as a combination of  $n$  basis vectors. The **standard basis** for  $\mathbb{R}^n$  is the set  $\{e_1, e_2, \dots, e_n\}$ . Other basis vectors are possible, but a potential set must contain  $n$  independent vectors in order to qualify as a basis.

As we have seen, for a matrix  $A$  there are two particularly interesting subspaces:

- The null space,  $\text{Nul } A$ , consisting of all solutions to  $A\vec{x} = 0$
- The column space,  $\text{Col } A$ , consisting of all linear combinations of the columns of  $A$ . If there is a solution to  $A\vec{x} = \vec{y}$  then  $\vec{y}$  must lie in the column space of  $A$  (i.e. it must be some linear combination of the columns of  $A$ ).

The dimension of  $\text{Nul } A$  is known as  $\text{nullity}(A)$ .

The dimension of  $\text{Col } A$  is  $\text{rank}(A)$ . In fact, the rank is also the dimension of the row space.

### Sylvester's Law of Nullity

For the  $m \times n$  matrix  $A$ :

$$\text{rank}(A) + \text{nullity}(A) = n$$

**Example:** Given a matrix  $A$  find a **basis** for its null space and the **dimension** of that basis.

**Answer:** We've done this!  $\text{Nul } A$ , can be found via reduced echelon form.

Previous example: Let

$$A = \begin{bmatrix} 1 & 2 & 3 & 1 \\ 1 & 3 & 5 & -2 \\ 3 & 8 & 13 & -3 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 0 & -1 & 7 \\ 0 & 1 & 2 & -3 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

$\text{Nul } A$  spanned by one vector for each free variable:

Here the free variables are  $x_3, x_4$ . First equation gives:

$$x_1 - x_3 + 7x_4 = 0 \implies x_1 = x_3 - 7x_4$$

Second equation gives:

$$x_2 + 2x_3 - 3x_4 = 0 \implies x_2 = -2x_3 + 3x_4$$

Combining:  $\vec{x} \in \text{Nul } A \iff$

$$\vec{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} x_3 - 7x_4 \\ -2x_3 + 3x_4 \\ x_3 \\ x_4 \end{bmatrix} = x_3 \begin{bmatrix} 1 \\ -2 \\ 1 \\ 0 \end{bmatrix} + x_4 \begin{bmatrix} -7 \\ 3 \\ 0 \\ 1 \end{bmatrix}$$

$$\text{Hence } \text{Nul } A = \text{Span} \left\{ \begin{bmatrix} 1 \\ -2 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} -7 \\ 3 \\ 0 \\ 1 \end{bmatrix} \right\}$$

The bottom rows show that these vectors are linearly independent, so they are a **basis** for  $\text{Nul } A$ . Since there are two,  $\text{nullity}(A) = 2$ .

**Example:** Given a matrix  $A$  find a **basis** for its column space and its **dimension**.

**Step 1:** Reduce to echelon form.

**Step 2:** Identify the **pivot columns**

**Step 3:** These columns of the **original matrix!** are a basis for  $\text{Col } A$ .

$$A = \begin{bmatrix} 1 & 2 & 3 & 1 \\ 1 & 3 & 5 & -2 \\ 3 & 8 & 13 & -3 \end{bmatrix} \xrightarrow{\text{row reduce}} \begin{bmatrix} 1 & 0 & -1 & 7 \\ 0 & 1 & 2 & -3 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

Pivot columns are first and second columns

First and second columns of **original  $A$**  span  $\text{Col } A$ :

$$\text{Col } A = \text{Span} \left\{ \begin{bmatrix} 1 \\ 1 \\ 3 \end{bmatrix}, \begin{bmatrix} 2 \\ 3 \\ 8 \end{bmatrix} \right\} \text{ These vectors form a } \mathbf{basis} \text{ for } \text{Col } A.$$

Note that Sylvester's Law of Nullity is satisfied:

$$\mathit{rank}(A) + \mathit{nullity}(A) = n$$