

# Some Subspaces

Ex. Solve the equation  $A\mathbf{x} = \mathbf{0}$  for  $A = \begin{bmatrix} -3 & 6 & -1 & 1 & -7 \\ 1 & -2 & 2 & 3 & -1 \\ 2 & -4 & 5 & 8 & -4 \end{bmatrix}$

The set of solutions to this system form a subspace because this set is the span of the vectors.

Def. The null space of matrix  $A$ , written  $\text{Nul } A$ , is the set of solutions to the homogeneous equation  $A\mathbf{x} = \mathbf{0}$ .

$$\{\mathbf{x} : A\mathbf{x} = \mathbf{0}\}$$

Note that this only works for the homogeneous equation.

→ The solution set for  $A\mathbf{x} = \mathbf{b}$  doesn't include the zero vector.

→ Also,  $A\mathbf{x} = \mathbf{b}$  may have no solution

Ex. For  $A = \begin{bmatrix} 1 & -3 & -2 \\ -5 & 9 & 1 \end{bmatrix}$ , determine if  $\mathbf{u} = \begin{bmatrix} 5 \\ 3 \\ -2 \end{bmatrix}$  is in the null space of  $A$ .

Another description:

Consider the linear transformation  $\mathbf{x} \mapsto A\mathbf{x}$ ,  
Nul  $A$  is the set of all vectors that are mapped  
to the zero vector.

There's no obvious relation between the entries of  $A$  and the vectors in  $\text{Nul } A$  (or its spanning set).

Another subspace, which has a more obvious connection, is the column space of  $A$ .

Def. The column space of  $A$ , written  $\text{Col } A$ , is the subspace that is the span of the columns of  $A$ .

If  $A = [\mathbf{a}_1 \quad \mathbf{a}_2 \quad \dots \quad \mathbf{a}_n]$ , a vector  $\mathbf{b}$  is in  $\text{Col } A$  if

$$\mathbf{b} = x_1 \mathbf{a}_1 + x_2 \mathbf{a}_2 + \dots + x_n \mathbf{a}_n$$

$$\{\mathbf{b} : A\mathbf{x} = \mathbf{b}\}$$

For the linear transformation  $\mathbf{x} \mapsto A\mathbf{x}$ ,  $\text{Col } A$  is the range.

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

$$W = \left\{ \begin{bmatrix} 6a - b \\ a + b \\ -7a \end{bmatrix} : a \text{ and } b \text{ are real numbers} \right\}$$

Ex. Consider the  $3 \times 4$  matrix  $A$ .

- a.  $\text{Col } A$  is a subspace of  $\mathbb{R}^k$  for what value of  $k$ ?
- b.  $\text{Nul } A$  is a subspace of  $\mathbb{R}^k$  for what value of  $k$ ?



Ex. For  $A = \begin{bmatrix} 1 & -3 & -2 \\ -5 & 9 & 1 \end{bmatrix}$ , determine if  $\mathbf{u} = \begin{bmatrix} 5 \\ 3 \end{bmatrix}$  is in the column space of  $A$ .

Ex. For  $A = \begin{bmatrix} 1 & -3 & -2 \\ -5 & 9 & 1 \end{bmatrix}$ , find a nonzero vector in Col  $A$

and a nonzero vector in Nul  $A$ .

Nul  $A$  and Col  $A$  are quite different, though we will find a connection between them next class.

### Contrast Between Nul $A$ and Col $A$ for an $m \times n$ Matrix $A$

Nul $A$	Col $A$
1. Nul $A$ is a subspace of $\mathbb{R}^n$ .	1. Col $A$ is a subspace of $\mathbb{R}^m$ .
2. Nul $A$ is implicitly defined; that is, you are given only a condition ( $Ax = \mathbf{0}$ ) that vectors in Nul $A$ must satisfy.	2. Col $A$ is explicitly defined; that is, you are told how to build vectors in Col $A$ .
3. It takes time to find vectors in Nul $A$ . Row operations on $[A \quad \mathbf{0}]$ are required.	3. It is easy to find vectors in Col $A$ . The columns of $A$ are displayed; others are formed from them.
4. There is no obvious relation between Nul $A$ and the entries in $A$ .	4. There is an obvious relation between Col $A$ and the entries in $A$ , since each column of $A$ is in Col $A$ .
5. A typical vector $\mathbf{v}$ in Nul $A$ has the property that $A\mathbf{v} = \mathbf{0}$ .	5. A typical vector $\mathbf{v}$ in Col $A$ has the property that the equation $Ax = \mathbf{v}$ is consistent.
6. Given a specific vector $\mathbf{v}$ , it is easy to tell if $\mathbf{v}$ is in Nul $A$ . Just compute $A\mathbf{v}$ .	6. Given a specific vector $\mathbf{v}$ , it may take time to tell if $\mathbf{v}$ is in Col $A$ . Row operations on $[A \quad \mathbf{v}]$ are required.
7. Nul $A = \{\mathbf{0}\}$ if and only if the equation $Ax = \mathbf{0}$ has only the trivial solution.	7. Col $A = \mathbb{R}^m$ if and only if the equation $Ax = \mathbf{b}$ has a solution for every $\mathbf{b}$ in $\mathbb{R}^m$ .
8. Nul $A = \{\mathbf{0}\}$ if and only if the linear transformation $\mathbf{x} \mapsto A\mathbf{x}$ is one-to-one.	8. Col $A = \mathbb{R}^m$ if and only if the linear transformation $\mathbf{x} \mapsto A\mathbf{x}$ maps $\mathbb{R}^n$ onto $\mathbb{R}^m$ .

When considering more abstract vector spaces, we discuss the linear transformation rather than the matrix.

Def. A linear transformation  $T$  from a vector space  $V$  to a vector space  $W$  is a rule that assigns to each vector  $\mathbf{x}$  in  $V$  a unique vector  $T(\mathbf{x})$  in  $W$ , such that

- i.  $T(\mathbf{u} + \mathbf{v}) = T(\mathbf{u}) + T(\mathbf{v})$
- ii.  $T(c\mathbf{u}) = cT(\mathbf{u})$

The kernel of  $T$  is the subspace of  $V$  that is mapped to the zero vector in  $W$ .

→ If  $T$  is a matrix transformation, this is the null space.

The range of  $T$  is the subspace of  $W$  of all vectors of the form  $T(\mathbf{x})$  for some  $\mathbf{x}$  in  $V$ .

→ If  $T$  is a matrix transformation, this is the column space.

Ex. An example of an abstract linear transformation would be the derivative.

We can use  $C[a,b]$ , which is the set of all continuous functions on the interval  $[a,b]$ .

Ex. Define the linear transformation  $T: \mathbf{P}_2 \rightarrow \mathbb{R}^2$

by  $T(\mathbf{p}) = \begin{bmatrix} \mathbf{p}(0) \\ \mathbf{p}'(0) \end{bmatrix}$ . Find the kernel of  $T$ .

# Linear Independence

A set of vectors  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_p$  is linearly dependent if there exist constants  $c_1, c_2, \dots, c_p$  (not all zero) such that

$$c_1\mathbf{v}_1 + c_2\mathbf{v}_2 + \dots + c_p\mathbf{v}_p = \mathbf{0}$$

- This equation is called a linear dependence relation.
- If the set is dependent, one of the vectors can be written as the linear combination of the others.
- The set is linearly independent if  $c_1 = c_2 = \dots = c_p = 0$  is the only solution.
- When we saw this before, the vectors were in  $\mathbb{R}^n$  and we looked at the equation  $A\mathbf{x} = \mathbf{0}$ .
- For abstract vector spaces, we can't rely on that.



Ex. In  $\mathbf{P}$ , determine if  $\mathbf{p}_1(t) = 1$ ,  $\mathbf{p}_2(t) = t$ ,  $\mathbf{p}_3(t) = t^2$ , and  $\mathbf{p}_4 = (t + 3)^2$  are linearly dependent.

Ex. In  $C[0,1]$ , determine if  $\{\cos t, \sin t\}$  is linearly dependent.

Ex. In  $C[0,1]$ , show that  $\{\cos t, \sin t, \sin(t + \frac{\pi}{4})\}$  is linearly dependent.

Def. Let  $H$  be a subspace of a vector space  $V$ . A set of vectors  $\mathcal{B}$  in  $V$  is a basis of  $H$  if

- i. The vectors in  $\mathcal{B}$  are linearly independent
- ii. The vectors in  $\mathcal{B}$  span  $H$ .

This could be considered the most “efficient” way to define the subspace  $H$ .

Ex. Determine if  $\mathbf{v}_1 = \begin{bmatrix} 3 \\ 0 \\ -6 \end{bmatrix}$ ,  $\mathbf{v}_2 = \begin{bmatrix} -4 \\ 1 \\ 7 \end{bmatrix}$ , and  $\mathbf{v}_3 = \begin{bmatrix} -2 \\ 1 \\ 5 \end{bmatrix}$  form a basis for  $\mathbb{R}^3$ .

The columns of  $I_n$  are called the standard basis for  $\mathbb{R}^n$ .

In  $\mathbb{R}^3$ , the standard basis vectors are

$$\mathbf{e}_1 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \mathbf{e}_2 = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \text{ and } \mathbf{e}_3 = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$$

The set  $S = \{1, t, t^2, \dots, t^n\}$  is called the standard basis for  $\mathbf{P}_n$ .

Ex. The vectors are dependent. If  $H = \text{span} \{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$ , identify a basis for  $H$ .

$$\mathbf{v}_1 = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}, \mathbf{v}_2 = \begin{bmatrix} 4 \\ 5 \\ 6 \end{bmatrix}, \mathbf{v}_3 = \begin{bmatrix} 2 \\ 1 \\ 0 \end{bmatrix}$$

A basis is a spanning set that is as small as possible.

Ex. Let  $H = \text{span}\{1, t, t^2, (t + 3)^2\}$ , find a basis.

Ex. Let  $H = \text{span}\{\cos t, \sin t\}$ , find a basis.

Ex. Let  $H = \text{span}\{\cos t, \sin t, \sin(t + \frac{\pi}{4})\}$ , find a basis.



We previously found vectors that span the null space of a vector  $A \rightarrow$  this will be the basis of  $\text{Nul } A$ .

Ex. Find a basis for  $\text{Nul } B$ , where

$$B = \begin{bmatrix} 1 & 4 & 0 & 2 & 0 \\ 0 & 0 & 1 & -1 & 0 \\ 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

It turns out that the pivot columns of a matrix form a basis for the column space of the matrix.

Ex. Find a basis for  $\text{Col } A$ , where

[This is row equiv. to  $B$ .]

$$A = \begin{bmatrix} 1 & 4 & 0 & 2 & -1 \\ 3 & 12 & 1 & 5 & 5 \\ 2 & 8 & 1 & 3 & 2 \\ 5 & 20 & 2 & 8 & 8 \end{bmatrix}$$

Be careful to use the columns of  $A$ , not the reduced form