


Section 4.2: Null & Column Spaces, Linear Transformations

Definition: Let A be an $m \times n$ matrix. The **null space** of A , denoted¹ by $\text{Nul } A$, is the set of all solutions of the homogeneous equation $A\mathbf{x} = \mathbf{0}$. That is

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

We can say that $\text{Nul } A$ is the subset of \mathbb{R}^n that gets mapped to the zero vector under the linear transformation $\mathbf{x} \mapsto A\mathbf{x}$.

¹Some authors will write $\text{Null}(A)$ —I tend to write two ells. 

Example

Determine Nul A where

$$A = \begin{bmatrix} 1 & 0 & 3 \\ 1 & 2 & 7 \end{bmatrix}.$$

The vectors \vec{x} in Nul A satisfy $A\vec{x} = \vec{0}$.

$$A \xrightarrow{\text{rref}} \begin{bmatrix} 1 & 0 & 3 \\ 0 & 1 & 2 \end{bmatrix}$$

$$x_1 = -3x_3$$

$$x_2 = -2x_3$$

x_3 -free

so $\vec{x} = \begin{bmatrix} -3x_3 \\ -2x_3 \\ x_3 \end{bmatrix} = x_3 \begin{bmatrix} -3 \\ -2 \\ 1 \end{bmatrix}$, we can say $\text{Nul } A = \text{Span} \left\{ \begin{bmatrix} -3 \\ -2 \\ 1 \end{bmatrix} \right\}$

Theorem

For $m \times n$ matrix A , $\text{Nul } A$ is a subspace of \mathbb{R}^n .

We can show that $\text{Nul } A$ contains the zero vector and is closed under the two operations.

Note, $A\vec{0} = \vec{0}$ hence $\vec{0}$ is in $\text{Nul } A$
in \mathbb{R}^n

If \vec{u} and \vec{v} are in $\text{Nul } A$, then $A\vec{u} = \vec{0}$ and $A\vec{v} = \vec{0}$.

So

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

$\vec{u} + \vec{v}$ solves the homogeneous equation, so $\vec{u} + \vec{v}$ is in $\text{Nul } A$. $\text{Nul } A$ is closed under vector addition.

Similarly, for scalar c

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

Hence $c\vec{u}$ is in $\text{Nul } A$ making it closed under scalar multiplication.

$\text{Nul } A$ is therefore a subspace of \mathbb{R}^n .

Example

For a given matrix, a spanning set for $\text{Nul}A$ gives an *explicit* description of this subspace. Find a spanning set for $\text{Nul} A$ where

$$A = \begin{bmatrix} 1 & 0 & 2 & -1 \\ 1 & 2 & 6 & -5 \end{bmatrix}.$$

Using the rref $A \xrightarrow{\text{rref}} \begin{bmatrix} 1 & 0 & 2 & -1 \\ 0 & 1 & 2 & -2 \end{bmatrix}$

$$x_1 = -2x_3 + x_4$$

$$x_2 = -2x_3 + 2x_4$$

$$x_3, x_4 - \text{free}$$

so for \vec{x} in $\text{Nul}A$

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

A explicit description of $\text{Nul } A$ is

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

Column Space

Definition: The **column space** of an $m \times n$ matrix A , denoted $\text{Col } A$, is the set of all linear combinations of the columns of A . If $A = [\mathbf{a}_1 \ \cdots \ \mathbf{a}_n]$, then

$$\text{Col } A = \text{Span}\{\mathbf{a}_1, \dots, \mathbf{a}_n\}.$$

Note that this corresponds to the set of \mathbf{b} for which linear equations of the form $A\mathbf{x} = \mathbf{b}$ are consistent! That is

$$\text{Col } A = \{\mathbf{b} \in \mathbb{R}^m \mid \mathbf{b} = A\mathbf{x} \text{ for some } \mathbf{x} \in \mathbb{R}^n\}.$$



Theorem

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

This follows from the previous theorem that says a span is always a subspace.

Corollary: $\text{Col } A = \mathbb{R}^m$ if and only if the equation $A\mathbf{x} = \mathbf{b}$ has a solution for every \mathbf{b} in \mathbb{R}^m .

Example

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

$$W = \left\{ \begin{bmatrix} 6a - b \\ a + b \\ -7a \end{bmatrix} \mid a, b \in \mathbb{R} \right\}.$$

We try to write elements of W as linear combinations.

$$\begin{bmatrix} 6a - b \\ a + b \\ -7a \end{bmatrix} = \begin{bmatrix} 6a \\ a \\ -7a \end{bmatrix} + \begin{bmatrix} -b \\ b \\ 0 \end{bmatrix} = a \begin{bmatrix} 6 \\ 1 \\ -7 \end{bmatrix} + b \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix}$$

$$= \begin{bmatrix} 6 & -1 \\ 1 & 1 \\ -7 & 0 \end{bmatrix} \begin{bmatrix} a \\ b \end{bmatrix}$$

$\underbrace{\hspace{10em}}$
A

$$A = \begin{bmatrix} 6 & -1 \\ 1 & 1 \\ -7 & 0 \end{bmatrix}$$

Example

$$A = \begin{bmatrix} 2 & 4 & -2 & 1 \\ -2 & -5 & 7 & 3 \\ 3 & 7 & -8 & 6 \end{bmatrix}$$

(a) If $\text{Col } A$ is a subspace of \mathbb{R}^k , what is k ?

Columns are in \mathbb{R}^3 so $k=3$

(b) If $\text{Nul } A$ is a subspace of \mathbb{R}^k , what is k ?

$A\vec{x}$ defined requires \vec{x} in \mathbb{R}^4
so $k=4$

$$A\vec{x} = \vec{0}$$

$3 \times 4 \quad k \times 1$

Example Continued...

$$A = \begin{bmatrix} 2 & 4 & -2 & 1 \\ -2 & -5 & 7 & 3 \\ 3 & 7 & -8 & 6 \end{bmatrix}, \quad \text{and} \quad \mathbf{u} = \begin{bmatrix} 3 \\ -2 \\ -1 \\ 0 \end{bmatrix}$$

(c) Is \mathbf{u} in $\text{Nul } A$? Could \mathbf{u} be in $\text{Col } A$?

↳ check: is $A\vec{u} = \vec{0}$? $A\vec{u} = \begin{bmatrix} 0 \\ -3 \\ 3 \end{bmatrix} \neq \vec{0}$

No, \vec{u} is not in $\text{Nul } A$.

\vec{u} is not in $\text{Col } A$. $\text{Col } A$ is a subspace of \mathbb{R}^3
while \vec{u} is in \mathbb{R}^4 .

Example Continued...

$$A = \begin{bmatrix} 2 & 4 & -2 & 1 \\ -2 & -5 & 7 & 3 \\ 3 & 7 & -8 & 6 \end{bmatrix}, \quad \text{and} \quad \mathbf{v} = \begin{bmatrix} 3 \\ -1 \\ 3 \end{bmatrix}$$

(c) Is \mathbf{v} in Col A ? Could \mathbf{v} be in Nul A ?

↓ Check: Is $A\vec{x} = \vec{v}$ consistent?

$$[A \ \vec{v}] \xrightarrow{\text{rref}} \begin{bmatrix} 1 & 0 & 9 & 0 & 5 \\ 0 & 1 & -5 & 0 & -30/17 \\ 0 & 0 & 0 & 1 & 1/17 \end{bmatrix}$$

← not a pivot column

$A\vec{x} = \vec{v}$ is consistent

Yes, \vec{v} is in Col A

$A\vec{v}$ isn't defined since \vec{v} is in \mathbb{R}^3 , so

no, \vec{v} could not be in $Nul A$.

Linear Transformation

Definition: Let V and W be vector spaces. A linear transformation $T : V \rightarrow W$ is a rule that assigns to each vector \mathbf{x} in V a unique vector $T(\mathbf{x})$ in W such that

- (i) $T(\mathbf{u} + \mathbf{v}) = T(\mathbf{u}) + T(\mathbf{v})$ for every \mathbf{u}, \mathbf{v} in V , and
- (ii) $T(c\mathbf{u}) = cT(\mathbf{u})$ for every \mathbf{u} in V and scalar c .

Example

Let $C^1(\mathbb{R})$ denote the set of all real valued functions that are differentiable and $C^0(\mathbb{R})$ the set of all continuous real valued functions. Note that differentiation is a linear transformation. That is

$$D : C^1(\mathbb{R}) \longrightarrow C^0(\mathbb{R}), \quad D(f) = f'$$

satisfies the two conditions in the previous definition.

From calculus, we know $\frac{d}{dx}(f(x)+g(x)) = f'(x) + g'(x)$

$$\text{i.e. } D(f+g) = D(f) + D(g)$$

$$\text{Also } \frac{d}{dx}(cf(x)) = cf'(x) \quad \text{i.e. } D(cf) = cD(f)$$

Characterize the subset of $C^1(\mathbb{R})$ such that $D(f) = 0$.

$D(f) = 0$ requires $f'(x) = 0$ for all x , $\Rightarrow f(x) = c$ for constant c .

Range and Kernel

Definition: The **range** of a linear transformation $T : V \longrightarrow W$ is the set of all vectors in W of the form $T(\mathbf{x})$ for some \mathbf{x} in V . (The set of all images of elements of V .)

Definition: The **kernel** of a linear transformation $T : V \longrightarrow W$ is the set of all vectors \mathbf{x} in V such that $T(\mathbf{x}) = \mathbf{0}$. (The analog of the null space of a matrix.)

Theorem: Given linear transformation $T : V \longrightarrow W$, the range of T is a subspace of W and the kernel of T is a subspace of V .

Example

Consider $T : C^1(\mathbb{R}) \longrightarrow C^0(\mathbb{R})$ defined by

$$T(f) = \frac{df}{dx} + \alpha f(x), \quad \alpha \text{ a fixed constant.}$$

(a) Express the equation that a function y must satisfy if y is in the kernel of T .

To be in the kernel $T(y) = 0$.

This requires

$$\frac{dy}{dx} + \alpha y = 0$$

Example

Consider $T : C^1(\mathbb{R}) \longrightarrow C^0(\mathbb{R})$ defined by

$$T(f) = \frac{df}{dx} + \alpha f(x), \quad \alpha \text{ a fixed constant.}$$

(b) Show that for any scalar c , $y = ce^{-\alpha x}$ is in the kernel of T .

y is the kernel if $\frac{dy}{dx} + \alpha y = 0$.

$$y = ce^{-\alpha x}, \quad \frac{dy}{dx} = -\alpha ce^{-\alpha x}$$

$$\text{so } \frac{dy}{dx} + \alpha y = -\alpha ce^{-\alpha x} + \alpha ce^{-\alpha x} = 0$$