

3.1 Image and Kernel

For a linear transformation $T: X \rightarrow Y$

the set X is the domain of T ,

the set Y is the codomain of T .

Definition The image of T is the

set $\text{image}(T) = \{T(\vec{x}) : \vec{x} \in X\}$.

$$= \{\vec{b} \in Y : \vec{b} = T(\vec{x}) \text{ for some } \vec{x} \in X\}$$

we might also call this the range of T

For example, suppose $T(\vec{y}) = (\vec{y})$. for $T: \mathbb{R}^2 \rightarrow \mathbb{R}^2$.

image $T = \{[\begin{smallmatrix} x \\ 0 \end{smallmatrix}]: x \in \mathbb{R}\}$.

$$T\left(\begin{bmatrix} x_1 \\ x_2 \end{bmatrix}\right) = \begin{bmatrix} 1 & 3 \\ 2 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = x_1 \begin{bmatrix} 1 \\ 2 \end{bmatrix} + x_2 \begin{bmatrix} 3 \\ 0 \end{bmatrix}$$

$$= x_1 \begin{bmatrix} 1 \\ 2 \end{bmatrix} + 3x_2 \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

$$= (x_1 + 3x_2) \begin{bmatrix} 1 \\ 2 \end{bmatrix}.$$

So every output of T is parallel to $\begin{bmatrix} 1 \\ 2 \end{bmatrix}$.

So $\text{image}(T) = \{c \begin{bmatrix} 1 \\ 2 \end{bmatrix} : c \in \mathbb{R}\}$.

Crucial Definition:

Given vectors $\vec{v}_1, \dots, \vec{v}_n$ in \mathbb{R}^k , the set of all linear combinations of $\vec{v}_1, \dots, \vec{v}_n$ is called the span and is denoted

$$\text{Span} \{ \vec{v}_1, \dots, \vec{v}_n \} = \{ c_1 \vec{v}_1 + \dots + c_n \vec{v}_n : c_1, \dots, c_n \in \mathbb{R} \}.$$

Lemma: The image of $T(\vec{x}) = A\vec{x}$ is the span of the column vectors of A .

Pf: Suppose $T(\vec{x}) = (\vec{v}_1 \ \vec{v}_2 \ \dots \ \vec{v}_n) \begin{pmatrix} x_1 \\ \vdots \\ x_m \end{pmatrix}$

$$= x_1 \vec{v}_1 + x_2 \vec{v}_2 + \dots + x_m \vec{v}_m$$

So every output of T is a linear combination of the columns of A and

Therefore: For $T: \mathbb{R}^n \rightarrow \mathbb{R}^m$ linear, $\text{im}(T)$ has these important properties.

① $T(\vec{0}) = \vec{0}$, so $\vec{0}$ is in $\text{im}(T)$.

② if \vec{b}, \vec{c} in $\text{im}(T)$,

$$\vec{b} = c_1 \vec{v}_1 + \dots + c_m \vec{v}_m \quad \text{for } c_1, \dots, c_m \text{ in } \mathbb{R}$$

$$\vec{c} = d_1 \vec{v}_1 + \dots + d_m \vec{v}_m \quad \text{for } d_1, \dots, d_m \text{ in } \mathbb{R}.$$

then $\vec{b} + \vec{c} = (c_1 \vec{v}_1 + \dots + c_m \vec{v}_m) + (d_1 \vec{v}_1 + \dots + d_m \vec{v}_m)$
 $= (c_1 + d_1) \vec{v}_1 + \dots + (c_m + d_m) \vec{v}_m$
in $\text{im}(T)$.

$\text{im}(T)$ is closed under vector addition

③ if \vec{b} in $\text{im}(T)$,

$$\vec{b} = c_1 \vec{v}_1 + \dots + c_m \vec{v}_m.$$

then $k\vec{b} = k(c_1 \vec{v}_1 + \dots + c_m \vec{v}_m)$
 $= (k c_1) \vec{v}_1 + \dots + (k c_m) \vec{v}_m.$
in $\text{im}(T)$

$\text{im}(T)$ is closed under scalar multiplication.

This will, later defined to be the properties that characterize subspaces.

The set of vectors that T sends to 0 is special.

Definition : the kernel (or null space) of

$T(\vec{x}) = A\vec{x}$ from \mathbb{R}^m to \mathbb{R}^n is the set

$$\{ \vec{x} \text{ in } \mathbb{R}^m : T(\vec{x}) = \vec{0} \}.$$

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

$\text{im}(T)$ lies in the codomain or target space

$\text{ker}(T)$ lies in the domain or initial space.

Example: Let $T_x = \begin{bmatrix} 1 & 2 & 3 \end{bmatrix} \vec{x}$.

Find $\text{im}(T)$ and $\text{ker}(T)$.

$$\text{im}(T) = \text{span} \{ [1], [2], [3] \}.$$

$$\text{ker}(T) = \text{solution to } \begin{bmatrix} 1 & 2 & 3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = [0].$$

$$\left[\begin{array}{ccc|c} 1 & 1 & 1 & 0 \\ 1 & 2 & 3 & 0 \end{array} \right]$$

$$\sim \left[\begin{array}{ccc|c} 1 & 1 & 1 & 0 \\ 0 & 1 & 2 & 0 \end{array} \right]$$

$$\sim \left[\begin{array}{cccc} 1 & 0 & -1 & 6 \\ 0 & 1 & 2 & 0 \end{array} \right]$$

$$x_1 = \alpha$$

$$x_2 = -2\alpha$$

$$x_3 = \alpha$$

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} \alpha \\ -2\alpha \\ \alpha \end{bmatrix} = \alpha \begin{bmatrix} 1 \\ -2 \\ 1 \end{bmatrix}$$

$$= \text{span} \left\{ \begin{bmatrix} 1 \\ -2 \\ 1 \end{bmatrix} \right\}.$$

Properties of kernel:

① $\vec{0}$ in $\text{ker}(T)$ ($T(\vec{0}) = \vec{0}$)

② if \vec{u}, \vec{v} in $\text{ker}(T)$, $\vec{u} + \vec{v}$ in $\text{ker}(T)$

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

③ if $\vec{u} \in \text{ker}(T)$,

$$T(k\vec{u}) = kT(\vec{u}) = k\vec{0} = \vec{0}$$

so $k\vec{u}$ in $\text{ker}T$.

How about invertible matrices?

If A is invertible,

$T(\vec{x}) = A\vec{x}$ has $\text{ker}(T)$ solution to

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

but in this case, the only solution is $\vec{0}$.

So if A is invertible, $\text{ker}(T) = \{\vec{0}\}$.

alternatively, every \vec{b} in \mathbb{R}^n has a solution to

$A\vec{x} = \vec{b}$, so every \vec{b} in $\text{im}(T)$.

$\text{im}(T) = \mathbb{R}^n \Leftrightarrow T$ is invertible.

A is $n \times n$

- A is invertible

- $A\vec{x} = \vec{b}$ has a unique solution for all \vec{b} .

- $A\vec{x} = \vec{0}$ is solved only by $\vec{x} = \vec{0}$

- $\text{rref}(A) = I_n$.

- $\text{rank}(A) = n$
- $\text{im}(A) = \mathbb{R}^n$
- $\text{ker}(A) = \{\vec{0}\}.$