

Lecture XXVIII: § 5.7 Linear Transformations

§ 5.8 Operations with linear transformations

Recall $T: V \longrightarrow W$
 linear transf $\rightsquigarrow \mathcal{N}(T) = \{ \vec{v} \in V : T(\vec{v}) = \vec{0}_W \}$ { subspace of V }
 (Nullspace of T)
 $\rightsquigarrow \mathcal{R}(T) = \{ \vec{w} \in W : \vec{w} = T(\vec{v}) \text{ for some } \vec{v} \in V \}$
 subspace of W (Range of T)

§1 One-To-One / Onto:

Def T is one-to-one (or injective) if $T(\vec{v}_1) = T(\vec{v}_2)$ implies $\vec{v}_1 = \vec{v}_2$

Def T is onto (or surjective) if every element of W comes from some element in V
 (i.e. $\mathcal{R}(T) = W$). Equivalently $\text{rank}(T) = \dim W$

Proposition T is one-to-one if and only if $\mathcal{N}(T) = \{ \vec{0}_V \}$ (nullity(T) = 0)

Reason: Assume T is one-to-one. If $\vec{v} \in \mathcal{N}(T)$, then $T(\vec{v}) = \vec{0}_W = T(\vec{0}_V)$

This forces $\vec{v} = \vec{0}_V$. We conclude $\mathcal{N}(T) = \{ \vec{0}_V \}$.

• Assume $\mathcal{N}(T) = \{ \vec{0}_V \}$ & pick \vec{v}_1, \vec{v}_2 with $T(\vec{v}_1) = T(\vec{v}_2)$. Then

$$\begin{aligned} \vec{0}_W &= T(\vec{v}_1) - T(\vec{v}_2) = T(\vec{v}_1) + (-1)T(\vec{v}_2) = T(\vec{v}_1) + T(-\vec{v}_2) \\ &= T(\vec{v}_1 - \vec{v}_2) \quad \text{so } \vec{v}_1 - \vec{v}_2 \in \mathcal{N}(T) = \{ \vec{0}_V \}. \end{aligned}$$

That means $\vec{v}_1 - \vec{v}_2 = \vec{0}$ so $\vec{v}_1 = \vec{v}_2$.

Key example: (Taking words) Fix V vector space with basis $\mathcal{B} = \{ \vec{v}_1, \dots, \vec{v}_p \}$

Let $T: V \longrightarrow \mathbb{R}^p$
 $\vec{v} \longmapsto [\vec{v}]_{\mathcal{B}}$ = words of \vec{v} relative to \mathcal{B}

(If $\vec{v} = a_1 \vec{v}_1 + a_2 \vec{v}_2 + \dots + a_p \vec{v}_p$, then $[\vec{v}]_{\mathcal{B}} = \begin{bmatrix} a_1 \\ a_2 \\ \vdots \\ a_p \end{bmatrix}$)

Prop: T is a linear transformation, 1-to-1 and onto.
 (last time)

• To show T is 1-to-1: it's enough to check $\mathcal{N}(T) = \{ \vec{0}_V \}$

By construction $[\vec{v}]_{\mathcal{B}} = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}$ means $\vec{v} = 0 \cdot \vec{v}_1 + 0 \cdot \vec{v}_2 + \dots + 0 \cdot \vec{v}_p$
 $= \vec{0}_V + \vec{0}_V + \dots + \vec{0}_V = \vec{0}_V$

so $\mathcal{N}(T) = \{\vec{0}_V\}$

• T skew T onto: $\begin{bmatrix} a_1 \\ \vdots \\ a_p \end{bmatrix} = T(\vec{v})$ where $\vec{v} = a_1 \vec{v}_1 + \dots + a_p \vec{v}_p$
So any vector in \mathbb{R}^p is in $\mathcal{R}(T)$.

§ 2. Invertible linear transformations

Fix $T: V \longrightarrow W$ a linear transformation

Definition: T is invertible (or an isomorphism) if T is both 1-to-1 and onto:

Reason: T is 1-to-1 and onto if, and only if, every \vec{w} in W comes from exactly one \vec{v} in V

In other words, for every \vec{w} in W there is a unique \vec{v} in V with $T(\vec{v}) = \vec{w}$

So, we can define a "return to sender" map $S: W \longrightarrow V$ where

$S(\vec{w}) = \vec{v}$ if \vec{v} is the unique vector in V with $T(\vec{v}) = \vec{w}$

It is clear that $S(T(\vec{v})) = \vec{v}$ for every \vec{v} in V
 $T(S(\vec{w})) = \vec{w}$ ——— \vec{w} in W

Proposition: S is a linear transformation

Proof: (1) If \vec{w}_1, \vec{w}_2 are in W , and $S(\vec{w}_1) = \vec{v}_1$, $S(\vec{w}_2) = \vec{v}_2$
(meaning $T(\vec{v}_1) = \vec{w}_1$ and $T(\vec{v}_2) = \vec{w}_2$) then

$$T(\vec{v}_1 + \vec{v}_2) = T(\vec{v}_1) + T(\vec{v}_2) = \vec{w}_1 + \vec{w}_2$$

This means that $S(\vec{w}_1 + \vec{w}_2) = \vec{v}_1 + \vec{v}_2 = S(\vec{w}_1) + S(\vec{w}_2)$. ✓

(2) $S(c\vec{w}) = cS(\vec{w})$ is checked similarly:

$S(\vec{w}) = \vec{v}$ means $T(\vec{v}) = \vec{w}$ so $T(c\vec{v}) = c\vec{w}$,

meaning $S(c\vec{w}) = c\vec{v} = cS(\vec{w})$ ✓

Example: ① $T: \mathbb{R} \rightarrow \mathbb{R}$ linear

$$x \mapsto 2x$$

$$S: \mathbb{R} \rightarrow \mathbb{R}$$

$$y \mapsto x \text{ with } T(x) = 2x = y \Rightarrow x = \frac{y}{2}$$

Conclusion: $S: \mathbb{R} \rightarrow \mathbb{R}$ is the inverse to T (it's linear)
 $y \mapsto \frac{y}{2}$

② $T: \mathbb{R}^n \rightarrow \mathbb{R}^n$ for a fixed $n \times n$ matrix A
 $\vec{x} \mapsto A\vec{x}$

Q: What can we say about A if T is invertible?

- 1-to-1: $\mathcal{N}(A) = \mathcal{N}(T) = \{0_V\}$, so A is non-singular
- onto: $\mathcal{R}(A) = \mathcal{R}(T) = \mathbb{R}^n \Rightarrow \text{rk } A = n$

Both conditions are equivalent to A being invertible

Conclusion T invertible means A is invertible

The inverse map $S: \mathbb{R}^n \rightarrow \mathbb{R}^n$ is given by
 $\vec{y} \mapsto \vec{x}$ with $A\vec{x} = \vec{y}$

multiplication by the $n \times n$ matrix A^{-1} (so it's a linear map!)

Indeed, $A\vec{x} = \vec{y}$ gives $A^{-1}(A\vec{x}) = A^{-1}\vec{y}$

$$(A^{-1}A)\vec{x} = A^{-1}\vec{y}$$

$$\vec{x} = I_n \vec{x} = A^{-1}\vec{y}$$

§3. Identity Transformation:

Fix a vector space V .

Definition: The identity transformation, denoted by $I_V: V \rightarrow V$, is the linear transformation $\text{Id}_V(\vec{v}) = \vec{v}$ for every \vec{v} in V .

Our calculations in §4 show:

Theorem: Fix vector spaces V, W & a linear transformation $T: V \rightarrow W$

The following assertions are equivalent to each other:

(1) T is both 1-to-1 and onto

(2) For every \vec{w} in W there is a unique \vec{v} in V with $T(\vec{v}) = \vec{w}$

(3) There is a linear transformation $S: W \rightarrow V$ with

$$\& S(T(\vec{v})) = \vec{v} \quad \text{for each } \vec{v} \text{ in } V \quad [S \circ T = \text{Id}_V]$$

$$\& T(S(\vec{w})) = \vec{w} \quad \text{for } \vec{w} \text{ in } W \quad [T \circ S = \text{Id}_W]$$

↳ "composition"

§4. Composition of Linear Transformations:

Assume we have 3 vector spaces and 2 linear transformations

$$T_1: U \rightarrow V \quad ; \quad T_2: V \rightarrow W$$

We can "compose" these just as with ordinary functions of 1 variable

$$T_2 \circ T_1: U \rightarrow W \quad (\text{read from right to left: apply } T_1 \text{ first and then } T_2)$$

$$T_2 \circ T_1(\vec{u}) = T_2(T_1(\vec{u})) \quad \text{for any } \vec{u} \text{ in } U$$

in V

Example:

$$T_1: \mathcal{P}_4 \rightarrow \mathcal{P}_3$$

$$f \longmapsto \frac{df}{dx}$$

linear

$$T_2: \mathcal{P}_3 \rightarrow \text{Mat}_{2 \times 3}$$

$$T_2(a + bx + cx^2 + dx^3) = \begin{bmatrix} a & 0 & b \\ c & d & 0 \end{bmatrix}$$

linear

$$\begin{aligned} T_2 \circ T_1(a_0 + a_1x + a_2x^2 + a_3x^3) &= T_2\left(\frac{d}{dx}(a_0 + a_1x + a_2x^2 + a_3x^3)\right) \\ &= T_2(\underbrace{a_1}_a + \underbrace{2a_2x}_b + \underbrace{3a_3x^2}_c + \underbrace{0}_{d=0}) = \begin{bmatrix} a_1 & 0 & 2a_2 \\ 3a_3 & 0 & 0 \end{bmatrix} \end{aligned}$$

