

vv214: Linear transformations II.

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This week

Today

1. Kernel and image of a linear transformation.
2. Rank-Nullity Theorem.
3. Inverse linear transformations.

Next class

Coordinates.

Image and Kernel of a Linear Transformation

Definition: The **kernel** and **image** of a linear operator $T: V \rightarrow W$ are defined by

$$\text{Ker } T = \{v \in V: Tv = 0\} \quad \text{Im } T = \{w \in W: w = Tv, v \in V\}$$

Examples: 1. $T: \mathbb{R} \rightarrow \mathbb{R}$, $Tx = x^2$ (not linear)

$$\text{Ker } T = \{0\}, \text{Im } T = \mathbb{R}_+ \cup \{0\}$$

2. $T: \mathbb{R} \rightarrow \mathbb{R}^2$, $T(t) = \begin{pmatrix} \cos t \\ \sin t \end{pmatrix}$ (not linear)

$$\text{Ker } T = \emptyset, \text{Im } T = \text{unit circle}$$

3. $T: \mathbb{R}^3 \rightarrow \mathbb{R}^3$, $T \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} x_1 \\ x_2 \\ 0 \end{pmatrix}$

$$\text{Ker } T = k \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}, k = \text{const}, \text{Im } T = xy \text{ plane}$$

Image and Kernel of a Linear Transformation

$$4. T: \mathbb{R}^2 \rightarrow \mathbb{R}^2, T\bar{x} = A\bar{x}, A = \begin{pmatrix} 2 & 3 \\ 6 & 9 \end{pmatrix}$$

$$\text{Ker } A = t \begin{pmatrix} -3/2 \\ 1 \end{pmatrix}, \text{Im } A = \text{span} \left(\begin{pmatrix} 1 \\ 3 \end{pmatrix} \right)$$

$$5. T: P_2(\mathbb{R}) \rightarrow P_2(\mathbb{R}), Tp(t) = p'(t)$$

$$p(t) = a_0 + a_1t + a_2t^2 \Rightarrow Tp(t) = p'(t) = a_1 + 2a_2t$$

$$A = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 0 \end{pmatrix}$$

$$\text{Ker } T = \{p(t): Tp = 0\} = \{a_0\} = \text{span}(1), \text{Im } T = \text{span}(1, t)$$

Image and Kernel of a Linear Transformation

Lemma 1: Let $T: \mathbb{R}^m \rightarrow \mathbb{R}^n$ be defined by the matrix $A_{n \times m}$. The columns of the matrix A are linearly independent iff

$$\text{Ker } A_{n \times m} = \{\bar{0}\} \iff \text{rank } A = m \Rightarrow m \leq n$$

Lemma 2: Let $T: V \rightarrow W$ be a linear operator. $\text{Im } T$ and $\text{Ker } T$ are linear subspaces of V and $W \Rightarrow$ there exist bases of the kernel and the image of a linear transformation.

Image and Kernel of a Linear Transformation

Definition: A map $T: V \rightarrow W$ is called **injective** if $Tu = Tv$ implies $u = v$.

"distinct inputs to distinct outputs"

Lemma: A linear operator $T: V \rightarrow W$ is injective iff $\text{Ker } T = \{0\}$.

- ▶ Let T be injective. As $\{0\} \subset \text{Ker } T$, so we need to show that $\text{Ker } T \subset \{0\}$.

$$\text{Let } v \in \text{Ker } T \Rightarrow Tv = 0 = T(0) \Rightarrow v = 0 \Rightarrow \text{Ker } T \subset \{0\}.$$

- ▶ Let $\text{Ker } T = \{0\}$. If $Tu = Tv$, then
$$T(u - v) = 0 \Rightarrow u - v \in \text{Ker } T \Rightarrow u - v = 0 \Rightarrow u = v$$

Example: $T: \mathbb{R}^2 \rightarrow \mathbb{R}^3$, $T(x, y) = (2x, 3y, x + 2y)$

Definition: A map $T: V \rightarrow W$ is called **surjective** if $\text{Im } T = W$.

Example: $T: P_5(\mathbb{R}) \rightarrow P_5(\mathbb{R})$, $Tp(t) = p'(t)$ is not surjective.

Image and Kernel of a Linear Transformation: Example

$$T: \mathbb{R}^6 \rightarrow \mathbb{R}^4, A = \begin{pmatrix} 0 & 1 & 2 & 0 & 0 & 3 \\ 0 & 0 & 0 & 1 & 0 & 4 \\ 0 & 0 & 0 & 0 & 1 & 5 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$

$$\Rightarrow \text{Ker } A = \{\bar{x} \in \mathbb{R}^6: A\bar{x} = 0\}$$

$$\Rightarrow \begin{cases} x_2 + 2x_3 + 3x_6 = 0 & \Rightarrow x_2 = -2x_3 - 3x_6 \\ x_4 + 4x_6 = 0 & \Rightarrow x_4 = -4x_6 \\ x_5 + 5x_6 = 0 & \Rightarrow x_5 = -5x_6 \end{cases}$$

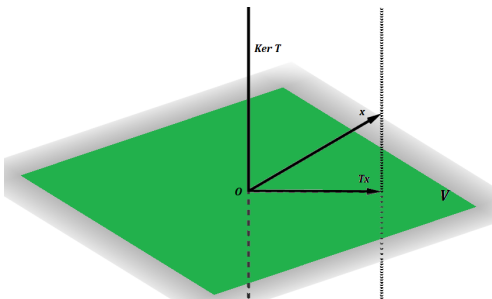
$$\bar{x} = r \begin{pmatrix} 1 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{pmatrix} + t \begin{pmatrix} 0 \\ -2 \\ 1 \\ 0 \\ 0 \\ 0 \end{pmatrix} + s \begin{pmatrix} 0 \\ -3 \\ 0 \\ -4 \\ -5 \\ 1 \end{pmatrix} \Rightarrow \dim \text{Ker } A = 3$$

Rank-Nullity Theorem

$$\dim \operatorname{Ker} T + \dim \operatorname{Im} T = \dim V$$

Example: $T: \mathbb{R}^3 \rightarrow \mathbb{R}^3$, $T\vec{x} = \operatorname{proj}_V \vec{x}$, $V \subset \mathbb{R}^3$

$$\operatorname{Ker} T = \{\vec{x} \in \mathbb{R}^3 : \operatorname{proj}_V \vec{x} = \vec{0}\}, \operatorname{Im} T = V$$



$\operatorname{Ker} T$ = line orthogonal to V

$$\underbrace{m}_3 - \underbrace{\dim(\operatorname{Ker} T)}_1 = \underbrace{\dim \operatorname{Im} T}_2$$

Rank-Nullity Theorem: Proof

Let $\dim(\text{Ker } T) = n$ and $\dim \text{Ker } T = k \Rightarrow k \leq n$.

\Rightarrow there exists a basis v_1, \dots, v_k , of $\text{Ker } T$. Complete this basis up to the basis of V : $v_1, \dots, v_k, v_{k+1}, \dots, v_n$

We are to prove that Tv_{k+1}, \dots, Tv_n form the basis for $\text{Im } T$:

1 Tv_{k+1}, \dots, Tv_n are linearly independent:

$$\alpha_1 Tv_{k+1} + \dots + \alpha_{n-k} Tv_n = 0 \Rightarrow T(\alpha_1 v_{k+1} + \dots + \alpha_{n-k} v_n) = 0$$

$$\Rightarrow \alpha_1 v_{k+1} + \dots + \alpha_{n-k} v_n \in \text{Ker } T$$

$$\Rightarrow \alpha_1 v_{k+1} + \dots + \alpha_{n-k} v_n \in \text{span}(v_1, \dots, v_k)$$

But $v_1, \dots, v_k, v_{k+1}, \dots, v_n$ are linearly independent

$$\Rightarrow \alpha_1 = \dots = \alpha_{n-k} = 0$$

2 $\text{span}(Tv_{k+1}, \dots, Tv_n) = \text{Im } T$

$$\text{A } w \in \text{Im } T \Rightarrow \exists v \in V: Tv = w \Rightarrow T(\beta_1 v_1 + \dots + \beta_n v_n) = w$$

$$w = \beta_1 \underbrace{Tv_1}_{=0} + \dots + \beta_k \underbrace{Tv_k}_{=0} + \beta_{k+1} Tv_{k+1} + \dots + \beta_n Tv_n$$

$$w \in \text{span}(Tv_{k+1}, \dots, Tv_n) \Rightarrow \text{Im } T \subset \text{span}(Tv_{k+1}, \dots, Tv_n)$$

$$\text{B } w \in \text{span}(Tv_{k+1}, \dots, Tv_n) \Rightarrow w = \alpha_{k+1} Tv_{k+1} + \dots + \alpha_{n-k} Tv_n$$

$$w = T(\alpha_{k+1} v_{k+1} + \dots + \alpha_{n-k} v_n) \Rightarrow w \in \text{Im } T$$

Inverse Linear Transformations

Definition: Let V, W be linear spaces.

A linear operator $T: V \rightarrow W$ is called **invertible** if there exists a linear operator $S: W \rightarrow V$ such that ST equals the identity map on V and TS equals the identity map on W .

A linear operator $S: W \rightarrow V$ satisfying $ST = I$ and $TS = I$ is called an **inverse** of T .

Here the first I is the identity map on V and the second I is the identity map on W . We shall denote the inverse linear operator by T^{-1} .

$$T^{-1}(Tv) = v \quad \text{and} \quad T(T^{-1}w) = w \quad \forall v \in V \forall w \in W$$

Inverse Linear Transformations

Lemma: A linear operator is invertible iff it is one-to-one (injective) and onto (surjective).

► Let T^{-1} exists.

A Let $u, v \in V$ and $Tu = Tv$

$$u = T^{-1}(Tu) = T^{-1}(Tv) = v \Rightarrow T \text{ is injective}$$

B Let $w \in W \Rightarrow w = T(T^{-1}w) \Rightarrow w \in \text{Im } T \Rightarrow W \subset \text{Im } T$

As also $\text{Im } T \subset W$, so $W = \text{Im } T$

► Let T be injective and surjective. For any $w \in W$, define Sw be a unique element of V such that $T(Sw) = w$. This element exists since T is one-to-one and onto.

A From the definition, $TS = I$. Also

$$T((ST)v) = (TS)(Tv) = ITv = Tv \Rightarrow STv = v \Rightarrow ST = I$$

B S is linear:

$$w_1, w_2 \in W \Rightarrow T(Sw_1 + Sw_2) = TS w_1 + TS w_2 = w_1 + w_2$$

Apply the definition of $S \Rightarrow S(w_1 + w_2) = Sw_1 + Sw_2$

Similarly, $S(\alpha w) = \alpha Sw \forall w \in W \forall \alpha \in \mathbb{K}$

Inverse Linear Transformations

Remarks:

1. $(T^{-1})^{-1} = T$
2. Let $V, W = \mathbb{R}^n$. A linear transformation $T: \mathbb{R}^n \rightarrow \mathbb{R}^n$ is invertible if the system $A\bar{x} = \bar{y}$ has a unique solution

$$\iff \text{rank } A = n \iff \text{rref } A = I_n$$

Definition: A square matrix A is invertible if the linear transformation $T\bar{x} = A\bar{x}$ is invertible.

Inverse Linear Transformations

$$A = \begin{pmatrix} 1 & 0 & -1 \\ 2 & 1 & 0 \\ -1 & 3 & 2 \end{pmatrix} \Rightarrow A\bar{x} = I\bar{y}$$

$$\begin{aligned} & \begin{pmatrix} 1 & 0 & -1 & 1 & 0 & 0 \\ 2 & 1 & 0 & 0 & 1 & 0 \\ -1 & 3 & 2 & 0 & 0 & 1 \end{pmatrix} \sim \begin{pmatrix} 1 & 0 & -1 & 1 & 0 & 0 \\ 0 & 1 & 2 & -2 & 1 & 0 \\ 0 & 3 & 1 & 1 & 0 & 1 \end{pmatrix} \\ & \sim \begin{pmatrix} 1 & 0 & -1 & 1 & 0 & 0 \\ 0 & 1 & 2 & -2 & 1 & 0 \\ 0 & 0 & -5 & 7 & -3 & 1 \end{pmatrix} \sim \begin{pmatrix} 1 & 0 & -1 & 1 & 0 & 0 \\ 0 & 1 & 2 & -2 & 1 & 0 \\ 0 & 0 & 1 & -\frac{7}{5} & \frac{3}{5} & -\frac{1}{5} \end{pmatrix} \\ & \sim \begin{pmatrix} 1 & 0 & -1 & 1 & 0 & 0 \\ 0 & 1 & 0 & \frac{4}{5} & -\frac{1}{5} & \frac{2}{5} \\ 0 & 0 & 1 & -\frac{7}{5} & \frac{3}{5} & -\frac{1}{5} \end{pmatrix} \sim \begin{pmatrix} 1 & 0 & 0 & -\frac{2}{5} & \frac{3}{5} & -\frac{1}{5} \\ 0 & 1 & 0 & \frac{4}{5} & -\frac{1}{5} & \frac{2}{5} \\ 0 & 0 & 1 & -\frac{7}{5} & \frac{3}{5} & -\frac{1}{5} \end{pmatrix} \\ & I\bar{y} = A^{-1}\bar{x} \Rightarrow A^{-1} = \frac{1}{5} \begin{pmatrix} -2 & 3 & -1 \\ 4 & -1 & 2 \\ -7 & 3 & -1 \end{pmatrix} \end{aligned}$$

Inverse Linear Transformations

1. Let $A_{n \times n}$. If A^{-1} exists, then the system $A\bar{x} = \bar{0}$ has a unique solution

$\Rightarrow \text{rank } A = n \Rightarrow$ columns of A are linearly independent.

2. If A^{-1} exists, then $A^{-1}A = AA^{-1} = I$.
3. $(AB)^{-1} = B^{-1}A^{-1}$