

Lecture 10: Linear Transformations and Subspaces

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Linear transformations are the primary functions between vector spaces that are of interest in linear algebra. They are special because they cooperate with the algebraic structure.

Definition Let m and n be positive integers. A linear transformation T from \mathbb{R}^n to \mathbb{R}^m is a function $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$ that satisfies

- $T(u + v) = T(u) + T(v)$, and
- $T(\lambda v) = \lambda T(v)$,

for all u and v in \mathbb{R}^n , and all scalars $\lambda \in \mathbb{R}$.

The Matrix of a Linear Transformation

Suppose that $T : \mathbb{R}^3 \rightarrow \mathbb{R}^2$ is a linear transformation, with

$$T \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} = \begin{bmatrix} 2 \\ -3 \end{bmatrix}, \quad T \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 1 \\ 4 \end{bmatrix}, \quad T \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} = \begin{bmatrix} -6 \\ 7 \end{bmatrix}$$

Then for the vector in \mathbb{R}^3 with any entries a, b, c

$$T \begin{bmatrix} a \\ b \\ c \end{bmatrix} = aT \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} + bT \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} + cT \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} = \underbrace{\begin{bmatrix} 2 & 1 & -6 \\ -3 & 4 & 7 \end{bmatrix}}_{M_T} \begin{bmatrix} a \\ b \\ c \end{bmatrix}.$$

and the 2×3 matrix M_T is called the (standard) matrix of T .

The matrix of a linear transformation

- A linear transformation $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is represented by a $m \times n$ matrix M_T . The **columns** of M_T are the images under T of the **standard basis vectors** e_1, \dots, e_n .
- If v is **any vector** in \mathbb{R}^n , we can calculate $T(v)$ by multiplying the column vector v on the left by the matrix M_T . **Matrix-vector multiplication is evaluating linear transformations.**
- On the other hand, if A is any $m \times n$ matrix, then A determines a linear transformation $\mathbb{R}^n \rightarrow \mathbb{R}^m$ by $v \rightarrow Av$, for $v \in \mathbb{R}^n$. So, in a sense, **matrices are linear transformations.**
- Examples of linear transformations include rotations, reflections and scaling, but not translations.
- If $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is a linear transformation, then in order to evaluate T at any point/vector, we only need mn pieces of information, just the m coordinates of each of the n images of the standard basis vectors. This is very different for example from continuous functions from \mathbb{R} to \mathbb{R} - we cannot know all about them just by knowing their values at a few points.

Matrix multiplication is composition

Suppose that $T : \mathbb{R}^n \rightarrow \mathbb{R}^p$ and $S : \mathbb{R}^p \rightarrow \mathbb{R}^m$ are linear transformations. Then $S \circ T$ (S after T) is the **linear transformation** from \mathbb{R}^n to \mathbb{R}^m defined for $v \in \mathbb{R}^n$ by

$$S \circ T(v) = S(T(v)).$$

Question How does the $(m \times n)$ matrix $M_{S \circ T}$ of $S \circ T$ depend on the $(m \times p)$ matrix M_S of S and the $(p \times n)$ matrix M_T of T ?

To answer this we have to think about the definition of $M_{S \circ T}$.

- Its first column has the coordinates of $S \circ T(e_1) = S(T(e_1))$.
- $T(e_1)$ is the first column of M_T .
- Then $S(T(e_1))$ is the matrix-vector product M_S [first column of M_T]. This is the first column of the matrix product $M_S M_T$.
- Same for all the other columns: the conclusion is $M_{S \circ T} = M_S M_T$.

Matrix multiplication is composition of linear transformations.

Corollary Matrix multiplication is associative.

The Image and Kernel of a Linear Transformation

$T : \mathbb{R}^3 \rightarrow \mathbb{R}^3$ is the linear transformation with $M_T = \begin{bmatrix} 1 & 2 & 0 \\ 2 & -1 & 5 \\ 1 & 1 & 1 \end{bmatrix}$.

The **image** of T is the subset of \mathbb{R}^3 consisting of all elements $T(v)$, where $v \in \mathbb{R}^3$. This is the set of all vectors of the form

$$a \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix} + b \begin{bmatrix} 2 \\ -1 \\ 1 \end{bmatrix} + c \begin{bmatrix} 0 \\ 5 \\ 1 \end{bmatrix}.$$

In matrix terms, this is the **column space** of M_T .

The **kernel** of T is the set of all vectors v in \mathbb{R}^3 with $T(v) = 0$.

This is the set of all column vectors whose entries a, b, c satisfy

$$\begin{bmatrix} 1 & 2 & 0 \\ 2 & -1 & 5 \\ 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} a \\ b \\ c \end{bmatrix} = a \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix} + b \begin{bmatrix} 2 \\ -1 \\ 1 \end{bmatrix} + c \begin{bmatrix} 0 \\ 5 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}.$$

In matrix terms this is the **(right) nullspace** of M_T .

Example: The kernel is a **line** and the image is a **plane**

$$\begin{bmatrix} 1 & 2 & 0 \\ 2 & -1 & 5 \\ 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} a \\ b \\ c \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \Rightarrow \begin{bmatrix} 1 & 2 & 0 & | & 0 \\ 2 & -1 & 5 & | & 0 \\ 1 & 1 & 1 & | & 0 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 0 & 2 & | & 0 \\ 0 & 1 & -1 & | & 0 \\ 0 & 0 & 0 & | & 0 \end{bmatrix}$$

The **kernel/nullspace** is $\boxed{\{(-2, 1, 1)t, t \in \mathbb{R}\}}$ a **line** in \mathbb{R}^3 .

That $(-2, 1, 1)$ is in the kernel of T means that (for example) Column 3 of M_T is a linear combination of Columns 1 and 2.

$$-2 \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix} + 1 \begin{bmatrix} 2 \\ -1 \\ 1 \end{bmatrix} + 1 \begin{bmatrix} 0 \\ 5 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \Rightarrow \begin{bmatrix} 0 \\ 5 \\ 1 \end{bmatrix} = 2 \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix} - \begin{bmatrix} 2 \\ -1 \\ 1 \end{bmatrix}$$

It follows that every linear combination of all three columns of M_T is actually a linear combination just of Columns 1 and 2.

The column space of M_T is $\left\{ a \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix} + b \begin{bmatrix} 2 \\ -1 \\ 1 \end{bmatrix} : a, b \in \mathbb{R} \right\}$, a **plane** in \mathbb{R}^3 .

Definition A (non-empty) subset V of \mathbb{R}^n is a **subspace** if

- It is **closed under addition**: $u + v \in V$ whenever $u \in V$ and $v \in V$.
- It is **closed under scalar multiplication**: $ku \in V$ whenever $u \in V$ and $k \in \mathbb{R}$.

Examples

- 1 $\{(x, y, z) \in \mathbb{R}^3 : x + y + z = 1\}$ is **not** a subspace of \mathbb{R}^3 . The vectors $(1, 0, 0)$ and $(0, 1, 0)$ belong to this set but their sum $(1, 1, 0)$ does not.
- 2 $\{(x, y, z) \in \mathbb{R}^3 : (x, y, z) \cdot (1, 2, 3) = 0\}$ **is** a subspace of \mathbb{R}^3 .
- 3 $\{(x, y, z) \in \mathbb{R}^3 : (x, y, z) \cdot (1, 2, 3) \neq 0\}$ is **not** a subspace of \mathbb{R}^3 . For example, $(1, 4, 1)$ and $(-5, -2, -1)$ belong to this set but their sum $(-4, 2, 0)$ does not.
- 4 The kernel of any linear transformation is a subspace.
- 5 The image of any linear transformation is a subspace.

Exercise Prove these last two points.