

Linear Transformations Between Vector Spaces

Linear Algebra
MATH 2076



Transformations $\mathbb{V} \xrightarrow{T} \mathbb{W}$

Let \mathbb{V} and \mathbb{W} be vector spaces, and let $\mathbb{V} \xrightarrow{T} \mathbb{W}$ be a function (aka, a *transformation*).

Here \mathbb{V} is the *domain* of T (where the input variables \vec{x} live) and \mathbb{W} is the *codomain* of T (where the resulting output $\vec{y} = T(\vec{x})$ lives).

For each \vec{x} in \mathbb{V} , $\vec{y} = T(\vec{x})$ is called the *image* of \vec{x} . If \mathbb{S} is a bunch of vectors in \mathbb{V} (i.e., \mathbb{S} is a *subset* of \mathbb{V}), then

$$T(\mathbb{S}) = \{ \text{all images } T(\vec{x}) \text{ where } \vec{x} \text{ is in } \mathbb{S} \}$$

is called the *T-image of \mathbb{S}* ; this is a subset of \mathbb{W} .

The *range* of T is the image of all of \mathbb{V} , i.e., $\boxed{\mathcal{Rng}(T) = T(\mathbb{V})}$; this is the set of all images $T(\vec{x})$, and $\mathcal{Rng}(T)$ is a subset of the codomain of T .

An important question is to know which vectors \vec{b} are in the range of T . That is: Given \vec{b} in \mathbb{W} , when can we find an \vec{x} in \mathbb{V} with $T(\vec{x}) = \vec{b}$?

Linear Transformations $\mathbb{V} \xrightarrow{T} \mathbb{W}$

Let \mathbb{V} and \mathbb{W} be vector spaces, and let $\mathbb{V} \xrightarrow{T} \mathbb{W}$ be a function (aka, a *transformation*).

We call T a *linear transformation* provided

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

and

$$T(s\vec{v}) = sT(\vec{v})$$

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

Note that for **any** LT T we *always* have $T(\vec{0}) = \vec{0}$. Right? Why?

Recall that every LT between two Euclidean spaces is a matrix transformation. In general?

Properties of Linear Transformations

Let $\mathbb{V} \xrightarrow{T} \mathbb{W}$ be a linear transformation. Then $T(\vec{0}) = \vec{0}$.

More importantly, T preserves all linear combinations; i.e., the T -image of a LC of vectors \vec{v}_j is a LC of $T(\vec{v}_j)$ using the same scalars. That is,

$$T(s_1\vec{v}_1 + s_2\vec{v}_2 + \cdots + s_p\vec{v}_p) = s_1 T(\vec{v}_1) + s_2 T(\vec{v}_2) + \cdots + s_p T(\vec{v}_p)$$

or more simply—using “summation” notation—

$$T\left(\sum_{j=1}^p s_j \vec{v}_j\right) = \sum_{j=1}^p s_j T(\vec{v}_j).$$

This is called the *linearity principle*.

Range and Kernel of a Linear Transformation

Let $\mathbb{V} \xrightarrow{T} \mathbb{W}$ be a linear transformation. The *range* of T is the image of all of \mathbb{V} , i.e.,

$$\mathcal{R}ng(T) = T(\mathbb{V}) = \{ T(\vec{v}) \mid \vec{v} \text{ in } \mathbb{V} \};$$

this is a vector subspace of \mathbb{W} .

The *kernel* of T is the pre-image of $\vec{0}$, i.e.,

$$\mathcal{K}er(T) = \{ \vec{v} \text{ in } \mathbb{V} \mid T(\vec{v}) = \vec{0} \};$$

this is a vector subspace of \mathbb{V} .

When \mathbb{V} and \mathbb{W} are Euclidean spaces, and T is mult by a matrix A

$$\mathcal{K}er(T) = \mathcal{N}S(A) \quad \text{and} \quad \mathcal{R}ng(T) = \mathcal{C}S(A).$$

Examples of Linear Transformations

Let A be a fixed $m \times n$ matrix. Define $\mathbb{R}^{n \times p} \xrightarrow{T} \mathbb{R}^{m \times p}$ by $T(X) = AX$; T is a linear transformation. Without knowing more about A , we cannot say much about $\mathcal{Ker}(T)$ or $\mathcal{Rng}(T)$.

We can also define $\mathbb{R}^{p \times m} \xrightarrow{S} \mathbb{R}^{p \times n}$ by $T(X) = XA$; S is an LT, but without knowing A , we can't "find" $\mathcal{Ker}(S)$ or $\mathcal{Rng}(S)$.

Now define $\mathbb{R}^{m \times n} \xrightarrow{T} \mathbb{R}^{n \times m}$ by $T(X) = X^T$; recall that X^T is the *transpose* of the matrix X given by $[X^T]_{ij} = [X]_{ji}$. Then T is an LT. What are $\mathcal{Ker}(T)$ and $\mathcal{Rng}(T)$? This is easy, right?

Next, define $\mathbb{R}^{n \times n} \xrightarrow{S} \mathbb{R}^{n \times n}$ by $T(X) = X - X^T$; you should check that S is an LT. What are $\mathcal{Ker}(S)$ and $\mathcal{Rng}(S)$?

If $S(X) = 0$, then $X = X^T$; so $\mathcal{Ker}(S)$ is the subspace of all *symmetric* $n \times n$ matrices.

HW: Show $\mathcal{Rng}(S)$ is the subspace of all *skew-symmetric* $n \times n$ matrices.

More Examples of Linear Transformations

Recall that \mathbb{P} is the vector space of all polynomials. Thus \mathbf{p} is in \mathbb{P} if and only if \mathbf{p} is a function of the form

$$\mathbf{p}(t) = c_0 + c_1t + c_2t^2 + \cdots + c_nt^n.$$

Here c_0, c_1, \dots, c_n are constants, called the *coefficients* of the polynomial \mathbf{p} , and when $c_n \neq 0$ we say that \mathbf{p} has *degree* n . (By definition, the zero polynomial has degree zero.)

Define $\mathbb{P} \xrightarrow{T} \mathbb{R}^3$ by $T(\mathbf{p}) = \begin{bmatrix} \mathbf{p}(1) \\ \mathbf{p}(2) \\ \mathbf{p}(3) \end{bmatrix}$. Check that T is an LT. What are $\mathcal{Ker}(T)$ & $\mathcal{Rng}(T)$?

This is not too hard; please try to find these yourself before we do it.

Clearly, $\mathcal{Rng}(T) = \mathbb{R}^3$. Right? Why?? Also, a poly \mathbf{p} is in $\mathcal{Ker}(T)$ iff \mathbf{p} has zeroes at each of $t = 1, 2, 3$; so \mathbf{p} is in $\mathcal{Ker}(T)$ iff $\deg(\mathbf{p}) \geq 3$ and \mathbf{p} can be factored as

$$\mathbf{p}(t) = (t - 1)(t - 2)(t - 3)\mathbf{q}(t).$$

More Examples of Linear Transformations

Recall that \mathbb{P} is the vector space of all polynomials. Define $\mathbb{P} \xrightarrow{D} \mathbb{P}$ by differentiation: $D(\mathbf{p}) = \mathbf{p}'$. By Calculus, D is an LT.

What are $\mathcal{Ker}(D)$ and $\mathcal{Rng}(D)$?

What is $\mathcal{Ker}(D)$? This is just the “trivial” subspace of all *constant* polynomials, right? So, $\mathcal{Ker}(D) = \mathcal{Span}\{\mathbf{1}\}$.

What is $\mathcal{Rng}(D)$? Which polys are the derivative of some other poly? Let \mathbf{q} be a poly, say

$$\mathbf{q}(t) = c_0 + c_1t + c_2t^2 + \cdots + c_nt^n.$$

Can we find some other poly \mathbf{p} so that $\mathbf{p}' = D(\mathbf{p}) = \mathbf{q}$? Sure! We just anti-differentiate, right? The poly

$$\mathbf{p}(t) = c_0t + \frac{c_1}{2}t^2 + \frac{c_2}{3}t^3 + \cdots + \frac{c_n}{n+1}t^{n+1}$$

has the property that $\mathbf{p}' = D(\mathbf{p}) = \mathbf{q}$. This means that $\mathcal{Rng}(D) = \mathbb{P}$.

And More Examples of Linear Transformations

Recall that \mathbb{P} is the vector space of all polynomials. Define $\mathbb{P}_n \xrightarrow{D} \mathbb{P}_n$ by differentiation: $D(\mathbf{p}) = \mathbf{p}'$. By Calculus, D is an LT.

This is clearly similar to the above example, but notice that here we are taking the domain and codomain of D to be \mathbb{P}_n instead of \mathbb{P} .

What are $\mathcal{Ker}(D)$ and $\mathcal{Rng}(D)$? This is not too hard; please try to find these yourself before we give the answers.

Again, $\mathcal{Ker}(D)$ is the “trivial” subspace of all *constant* polynomials, right?

What about $\mathcal{Rng}(D)$?

HW: Show that $\mathcal{Rng}(D) = \mathbb{P}_{n-1}$.

One More Example of a Linear Transformation

Recall that \mathbb{P} is the vector space of all polynomials. Define $\mathbb{P} \xrightarrow{S} \mathbb{P}$ by “multiplication”: i.e., $\mathbf{q} = S(\mathbf{p})$ is the polynomial given by $\mathbf{q}(t) = t \mathbf{p}(t)$. You should check that S is an LT.

What are $\mathcal{Ker}(S)$ and $\mathcal{Rng}(S)$?

Remember, $\mathcal{Ker}(S)$ is all polys \mathbf{p} with $S(\mathbf{p}) = \mathbf{0}$. Notice that the poly $\mathbf{1}$ has $S(\mathbf{1}) = \mathbf{t}$, right? So, every constant non-zero poly has a non-zero S image. We now see that $\mathcal{Ker}(S) = \{\mathbf{0}\}$.

What about $\mathcal{Rng}(S)$? Notice that every S image $\mathbf{q} = S(\mathbf{p})$ has $\mathbf{q}(0) = 0$. Using this we see that $\mathcal{Rng}(S)$ consists of all polys \mathbf{q} with “no constant term”; i.e.,

$$\mathcal{Rng}(S) = \left\{ \mathbf{q} \mid \mathbf{q}(t) = b_1 t + b_2 t^2 + \cdots + b_n t^n \right\}.$$

HW: Show that $DS - SD = I$ where D is “differentiation” (as described 2 slides above) and I is the identity transformation $I(\mathbf{p}) = \mathbf{p}$.