

## 23. Kernel, Rank, Range

We now study linear transformations in more detail. First, we establish some important vocabulary.

The *range* of a linear transformation  $f : V \rightarrow W$  is the set of vectors the linear transformation maps to. This set is also often called the *image* of  $f$ , written

$$\text{ran}(f) = \text{Im}(f) = L(V) = \{L(v) | v \in V\} \subset W.$$

The *domain* of a linear transformation is often called the *pre-image* of  $f$ . We can also talk about the pre-image of any subset of vectors  $U \subset W$ :

$$L^{-1}(U) = \{v \in V | L(v) \in U\} \subset V.$$

A linear transformation  $f$  is *one-to-one* if for any  $x \neq y \in V$ ,  $f(x) \neq f(y)$ . In other words, different vector in  $V$  always map to different vectors in  $W$ . One-to-one transformations are also known as *injective* transformations. Notice that injectivity is a condition on the pre-image of  $f$ .

A linear transformation  $f$  is *onto* if for every  $w \in W$ , there exists an  $x \in V$  such that  $f(x) = w$ . In other words, every vector in  $W$  is the image of some vector in  $V$ . An onto transformation is also known as an *surjective* transformation. Notice that surjectivity is a condition on the image of  $f$ .<sup>1</sup>

Suppose  $L : V \rightarrow W$  is *not* injective. Then we can find  $v_1 \neq v_2$  such that  $Lv_1 = Lv_2$ . Then  $v_1 - v_2 \neq 0$ , but

$$L(v_1 - v_2) = 0.$$

**Definition** Let  $L : V \rightarrow W$  be a linear transformation. The set of all vectors  $v$  such that  $Lv = 0_W$  is called the *kernel* of  $L$ :

$$\ker L = \{v \in V | Lv = 0\}.$$

---

<sup>1</sup> The notions of one-to-one and onto can be generalized to arbitrary functions on sets. For example if  $g$  is a function from a set  $S$  to a set  $T$ , then  $g$  is one-to-one if different objects in  $S$  always map to different objects in  $T$ . For a linear transformation  $f$ , these sets  $S$  and  $T$  are then just vector spaces, and we require that  $f$  is a linear map; *i.e.*  $f$  respects the linear structure of the vector spaces.

The linear structure of sets of vectors lets us say much more about one-to-one and onto functions than one can say about functions on general sets. For example, we always know that a linear function sends  $0_V$  to  $0_W$ . Then we can show that a linear transformation is one-to-one if and only if  $0_V$  is the only vector that is sent to  $0_W$ : by looking at just one (very special) vector, we can figure out whether  $f$  is one-to-one. For arbitrary functions between arbitrary sets, things aren't nearly so convenient!

**Theorem.** A linear transformation  $L$  is injective if and only if  $\ker L = \{0_V\}$ .

The proof of this theorem is an exercise.

Notice that if  $L$  has matrix  $M$  in some basis, then finding the kernel of  $L$  is equivalent to solving the homogeneous system

$$MX = 0.$$

**Example** Let  $L(x, y) = (x + y, x + 2y, y)$ . Is  $L$  one-to-one?

To see, we can solve the linear system:

$$\left( \begin{array}{cc|c} 1 & 1 & 0 \\ 1 & 2 & 0 \\ 0 & 1 & 0 \end{array} \right) \sim \left( \begin{array}{cc|c} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{array} \right).$$

Then all solutions of  $MX = 0$  are of the form  $x = y = 0$ . In other words,  $\ker L = 0$ , and so  $L$  is injective.

**Theorem.** Let  $L : V \rightarrow W$ . Then  $\ker L$  is a subspace of  $V$ .

*Proof.* Notice that if  $L(v) = 0$  and  $L(u) = 0$ , then for any constants  $c, d$ ,  $L(cu + dv) = 0$ . Then by the subspace theorem, the kernel of  $L$  is a subspace of  $V$ .  $\square$

This theorem has a nice interpretation in terms of the eigenspaces of  $L$ . Suppose  $L$  has a zero eigenvalue. Then the associated eigenspace consists of all vectors  $v$  such that  $Lv = 0v = 0$ ; in other words, the 0-eigenspace of  $L$  is exactly the kernel of  $L$ .

Returning to the previous example, let  $L(x, y) = (x + y, x + 2y, y)$ .  $L$  is clearly not surjective, since  $L$  sends  $\mathbb{R}^2$  to a plane in  $\mathbb{R}^3$ .

Notice that if  $x = L(v)$  and  $y = L(u)$ , then for any constants  $c, d$  then  $cx + dy = L(cx + du)$ . Then the subspace theorem strikes again, and we have the following theorem.

**Theorem.** Let  $L : V \rightarrow W$ . Then the image  $L(V)$  is a subspace of  $W$ .

To find a basis of the the image of  $L$ , we can start with a basis  $S = \{v_1, \dots, v_n\}$  for  $V$ , and conclude that

$$L(V) = \text{span } L(S) = \text{span}\{v_1, \dots, v_n\}.$$

However, the set  $\{v_1, \dots, v_n\}$  may not be linearly independent, so we solve

$$c^1 L(v_1) + \dots + c^n L(v_n) = 0.$$

By finding relations amongst  $L(S)$ , we can discard vectors until a basis is arrived at. The size of this basis is the dimension of the image of  $L$ , which is known as the *rank* of  $L$ .

**Definition** The *rank* of a linear transformation  $L$  is the dimension of its image, written  $\text{rank } L$ .

The *nullity* of a linear transformation is the dimension of the kernel, written  $\text{nullity } L$ .

**Theorem** (Dimension Formula). *Let  $L : V \rightarrow W$  be a linear transformation, with  $V$  a finite-dimensional vector space<sup>2</sup>. Then:*

$$\begin{aligned}\dim V &= \dim \ker L + \dim \text{Im } L \\ &= \dim \ker L + \text{rank } L.\end{aligned}$$

*Proof.* Pick a basis for  $V$ :

$$\{v_1, \dots, v_p, u_1, \dots, u_q\},$$

where  $v_1, \dots, v_p$  is also a basis for  $\ker L$ . This can always be done, for example, by finding a basis for the 0-eigenspace of  $L$ . Then  $p = \dim \ker L$  and  $p + q = \dim V$ . Then we need to show that  $q = \text{rank } L$ . To accomplish this, we show that  $\{L(u_1), \dots, L(u_q)\}$  is a basis for  $\text{Im } L$ .

To see that  $\{L(u_1), \dots, L(u_q)\}$  spans  $\text{Im } L$ , consider any vector  $w$  in  $\text{Im } L$  and find constants  $c^i, d^j$  such that:

$$\begin{aligned}w &= L(c^1 v_1 + \dots + c^p v_p + d^1 u_1 + \dots + d^q u_q) \\ &= c^1 L(v_1) + \dots + c^p L(v_p) + d^1 L(u_1) + \dots + d^q L(u_q) \\ &= d^1 L(u_1) + \dots + d^q L(u_q) \text{ since } L(v_i) = 0,\end{aligned}$$

$$\Rightarrow \text{Im } L = \text{span}\{L(u_1), \dots, L(u_q)\}.$$

Now we show that  $\{L(u_1), \dots, L(u_q)\}$  is linearly independent. We argue by contradiction: Suppose there exist constants  $c^j$  (not all zero) such that

$$\begin{aligned}0 &= d^1 L(u_1) + \dots + d^q L(u_q) \\ &= L(d^1 u_1 + \dots + d^q u_q).\end{aligned}$$

---

<sup>2</sup>The formula still makes sense for infinite dimensional vector spaces, such as the space of all polynomials, but the notion of a basis for an infinite dimensional space is more sticky than in the finite-dimensional case. Furthermore, the dimension formula for infinite dimensional vector spaces isn't useful for computing the rank of a linear transformation, since an equation like  $\infty = 3 + x$  cannot be solved for  $x$ . As such, the proof presented assumes a finite basis for  $V$ .

But since the  $u^j$  are linearly independent, then  $d^1u_1 + \dots + d^qu_q \neq 0$ , and so  $d^1u_1 + \dots + d^qu_q$  is in the kernel of  $L$ . But then  $d^1u_1 + \dots + d^qu_q$  must be in the span of  $\{v_1, \dots, v_p\}$ , since this was a basis for the kernel. This contradicts the assumption that  $\{v_1, \dots, v_p, u_1, \dots, u_q\}$  was a basis for  $V$ , so we are done.  $\square$

## References

- Hefferon, Chapter Three, Section II.2: Rangespace and Nullspace (Recall that ‘homomorphism’ is used instead of ‘linear transformation’ in Hefferon.)

Wikipedia:

- Rank
- Dimension Theorem
- Kernel of a Linear Operator

## Review Questions

1. Let  $L : V \rightarrow W$  be a linear transformation. Prove that  $\ker L = \{0_V\}$  if and only if  $L$  is one-to-one.
2. Let  $\{v_1, \dots, v_n\}$  be a basis for  $V$ . Explain why

$$L(V) = \text{span}\{L(v_1), \dots, L(v_n)\}.$$

3. Suppose  $L : \mathbb{R}^4 \rightarrow \mathbb{R}^4$  whose matrix  $M$  in the standard basis is row equivalent to the following matrix:

$$\begin{pmatrix} 1 & 0 & 0 & -1 \\ 0 & 1 & 0 & 1 \\ 0 & 0 & 1 & 1 \end{pmatrix}.$$

*Explain* why the first three columns of the original matrix  $M$  form a basis for  $L(V)$ .

*Find and describe* an algorithm (*i.e.* a general procedure) for finding a basis for  $L(V)$  when  $L : \mathbb{R}^n \rightarrow \mathbb{R}^m$ .

Finally, provide an example of the use of your algorithm.

4. Claim: If  $\{v_1, \dots, v_n\}$  is a basis for  $\ker L$ , where  $L : V \rightarrow W$ , then it is always possible to extend this set to a basis for  $V$ .

Choose a simple yet non-trivial linear transformation with a non-trivial kernel and verify the above claim for the transformation you choose.

5. Let  $P_n(x)$  be the space of polynomials in  $x$  of degree less than or equal to  $n$ , and consider the derivative operator  $\frac{d}{dx}$ . Find the dimension of the kernel and image of  $\frac{d}{dx}$ .

Now, consider  $P_2(x, y)$ , the space of degree two polynomials in  $x$  and  $y$ . (Recall that  $xy$  is degree two, and  $x^2y$  is degree three, for example.) Let  $L = \frac{d}{dx} + \frac{d}{dy}$ . (For example,  $L(xy) = \frac{d}{dx}(xy) + \frac{d}{dy}(xy) = y + x$ .) Find a basis for the kernel of  $L$ . Verify the dimension formula in this case.