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1. INJECTIONS, SURJECTIONS, AND ISOMORPHISMS

Definition 1. A linear transformation $T: \mathbb{R}^n \to \mathbb{R}^m$ is called:

• Injective, or one-to-one if for any vector $\mathbf{b} \in \mathbb{R}^m$ there is at most one vector $\mathbf{x} \in \mathbb{R}^n$ with $T(\mathbf{x}) = \mathbf{b}$. Equivalently if

$$T(\mathbf{x}) = T(\mathbf{y}) \Longrightarrow \mathbf{x} = \mathbf{y}$$

- Surjective, or onto if for any vector $\mathbf{b} \in \mathbb{R}^m$ there is at least one vector $\mathbf{x} \in \mathbb{R}^n$ with $T(\mathbf{x}) = \mathbf{b}$.
- an isomorphism if it's both injective and surjective. In other words, if for any vector $\mathbf{b} \in \mathbb{R}^m$ there is exactly one vector $\mathbf{x} \in \mathbb{R}^n$ with $T(\mathbf{x}) = \mathbf{b}$.

Example 2. Consider the following linear transformation $T: \mathbb{R}^2 \to \mathbb{R}$:

$$(x,y) \mapsto x-y$$

This is a surjective linear transformation. Indeed given any real number (say 3) we can find a vector $\mathbf{x} = (x, y) \in \mathbb{R}^2$ so that $T(\mathbf{x}) = 3$. For example we can take (x, y) = (1, -2). In fact, there is a whole line in \mathbb{R}^2 , the line with equation x - y = 3, whose each point is mapped to 5. Of course, there is nothing special about 3, for any real number b there is a whole line of points (x, y) so that T((x, y)) = b.

This means that T is *not* injective. Many vectors of \mathbb{R}^2 are mapped to the same vector in \mathbb{R} .

Example 3. Consider now the linear transformation $T: \mathbb{R}^4 \to \mathbb{R}^2$ given by $T(\mathbf{x}) = A\mathbf{x}$ where A is the following matrix:

$$A = \begin{pmatrix} 1 & 2 & 1 & -3 \\ 3 & -2 & 3 & -1 \end{pmatrix}$$

in other words $T: (x_1, x_2, x_3, x_4) \mapsto (x_1 + 2x_2 + x_3 - 3x_4, 3x_1 - 2x_2 + 3x_3 - x_4).$

Is this transformation surjective? This question translates to: "Does the system $A\mathbf{x} = \mathbf{b}$ has at least one solution for all $\mathbf{b} \in \mathbb{R}^2$?. If $\mathbf{b} = (b_1, b_2)$ then the augmented matrix of the system is

$$\begin{pmatrix} 1 & 2 & 1 & -3 & b_1 \\ 3 & -2 & 3 & -1 & b_2 \end{pmatrix}$$

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whose RREF is:

(1)
$$\begin{pmatrix} \boxed{1} & 0 & 1 & -1 & \frac{b_1 + b_2}{4} \\ 0 & \boxed{1} & 0 & -1 & \frac{3b_1 - b_2}{8} \end{pmatrix}$$

So the system has solutions for every \mathbf{b} and the transformation is injective. Since there are non-pivot columns in the RREF the solutions are not unique, and so the transformation is not injective.

Example 4. Consider the linear transformation $S: \mathbb{R}^2 \to \mathbb{R}^3$ whose matrix is

$$[S] = \begin{pmatrix} -2 & 11\\ 3 & -13\\ 1 & -5 \end{pmatrix}$$

Look at the system $[T](\mathbf{x}) = \mathbf{b}$. The augmented matrix is

$$\begin{pmatrix} -2 & 11 & b_1 \\ 3 & -13 & b_2 \\ 1 & -5 & b_3 \end{pmatrix}$$

Te RREF is:

(2)
$$\begin{pmatrix} 1 & 0 & 5b_1 + 11b_3 \\ 0 & 1 & b_1 + 2b_3 \\ 0 & 0 & -2b_1 + b_2 - 7b_3 \end{pmatrix}$$

The third row tell us that in order for the system to be consistent we need

$$(3) -2b_1+b_2-7b_3=0$$

So this transformation is not surjective.

On the other hand as long as Condition (3) is satisfied the system has a unique solution. So this transformation is injective.

We now define two important subspaces associated with a linear transformation.

Definition 5. Let $T: \mathbb{R}^n \to \mathbb{R}^m$ be a linear transformation.

• The *kernel* or *null space* of T is the subspace of \mathbb{R}^n that consists of all vectors that are mapped to **0**:

$$\ker T = \{ \mathbf{x} \in \mathbb{R}^n : T(\mathbf{x}) = \mathbf{0} \}$$

The notation $\mathcal{N}(T)$ is often also used for this subspace.

• The range or image of T is the subspace of \mathbb{R}^m that consists of all vectors **b** so that the system $T(\mathbf{x}) = \mathbf{b}$ has a solution:

im
$$T = \{ \mathbf{b} \in \mathbb{R}^m : T(\mathbf{x}) = \mathbf{b}, \text{ for some } \mathbf{x} \in \mathbb{R}^n \}$$

The notation $\mathcal{R}(T)$ is also often used for this subspace.

The following holds:

Proposition 6. Let $T: \mathbb{R}^n \to \mathbb{R}^m$ be a linear transformation.

• T is surjective if and only if, im $T = \mathbb{R}^m$

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• T is injective if and only if, $\ker T = \{\mathbf{0}\}$

Proof. The first assertion is obvious, so let's just prove the second.

The only if part: Clearly 0 is always in the kernel of T since T(0) = 0. If T is injective there can be no other vector \mathbf{x} with $T(\mathbf{x}) = \mathbf{0}$.

The if part: Assume ker $T = \{0\}$. We prove that then T is injective: let $\mathbf{x}, \mathbf{y} \in \mathbb{R}^n$, we have

$$T(\mathbf{x}) = T(\mathbf{y}) \Longrightarrow T(\mathbf{x}) - T(\mathbf{y}) = \mathbf{0}$$
$$\implies T(\mathbf{x} - \mathbf{y}) = \mathbf{0}$$
$$\implies \mathbf{x} - \mathbf{y} \in \ker T$$
$$\implies \mathbf{x} - \mathbf{y} = \mathbf{0}$$
$$\implies \mathbf{x} = \mathbf{y}$$

Example 7. Let's revisit Example 2. Since T is surjective, we have im $T = \mathbb{R}$.

To find the kernel of T we need to look at the set of vectors that are mapped to the zero vector. In other words look at the solution of the linear system

$$T(\mathbf{x}) = 0$$

This is the line K with equation x - y = 0. Now the set of solutions to this equation is described in parametric form as:

$$\mathbf{x} = s(1,1), \quad s \in \mathbf{R}$$

In other words

$$\ker T = \langle (1,1) \rangle$$

Since $\{(1,1)\}$ is a linearly independent set, it follows that $\{(1,1)\}$ is a basis of ker T so ker T is one-dimensional.

Now when we look at the set of vectors that are mapped to any other vector in \mathbb{R} , say 3, we get again a line, in this case the line L_3 : x - y = 3, and this line is a translation of the line K. In fact L_3 is the translation of L_3 by the vector (1, -2) that is a particular solution of the equation x - y = 3. The same happens with L_{-2} the set of vectors that are mapped to -2, L_{-2} is a translation of K by the vector, (0, -2) that is a particular solution of x - y = -2.

In general, the set L_b of vectors that are mapped to $b \in \mathbb{R}$ is a translation of K by any particular solution, for example we could take the particular solution $\mathbf{p}_b = (0, b)$. Then we can write

$$L_b = \mathbf{p}_b + K := \{\mathbf{p}_b + \mathbf{x} : \mathbf{x} + K\}$$

Example 8. Let's now look at Example 3. Again since T is surjective we have that im $T = \mathbb{R}^2$.



FIGURE 1. The transformation T of Example 2

To find the kernel we have to solve the system $[T](\mathbf{x}) = \mathbf{0}$. From the RREF of its augmented matrix we see that x_3 and x_4 are free variables, so by setting $x_3 = s$ and $x_4 = s$ we get the parametric form:

$$(x_1, x_2, x_3, x_4) = (s - t, -t, s, t)$$

or equivalently:

$$\mathbf{x} = s(1,0,1,0) + t(-1,-1,0,t), \quad s,t \in \mathbb{R}$$

In other words

$$\ker T = \langle (1, 0, 1, 0), (-1, -1, 0, 1) \rangle$$

and since $\{(1, 0, 1, 0), (-1, -1, 0, 1)\}$ is a linear independent set it is a basis of ker T. So the kernel is 2-dimensional.

Notice that the non-pivot columns of the RREF gave us a basis of ker T by the above procedure. Each non-pivot column gives a free variable, and then a vector in the basis of ker T.

What about the set $L_{\mathbf{b}}$ of vectors that are mapped to a vector \mathbf{b} , that is the solutions of the system $[T]\mathbf{x} = \mathbf{b}$?

From (1) we see that

$$L_{\mathbf{b}} = \mathbf{p}_{\mathbf{b}} + \ker T$$

where, $\mathbf{p_b} = (\frac{b_1+b_2}{4}, \frac{3b_1-b_2}{8})$. Of course, $\mathbf{p_b}$ is a particular solution.

In general we have that

Proposition 9. For a linear transformation $T: \mathbf{R}^n \to \mathbf{R}^m$, and a vector $\mathbf{b} \in \mathbb{R}^M$, the set of vectors $L_{\mathbf{b}}$ that are mapped to \mathbf{b} , is a translation of ker T by a (any) particular vector in $L_{\mathbf{b}}$.

Example 10. Let's look at Example 4. Now we have an injective transformation, so ker $T = \{0\}$.

To find im T we have to find for which $\in \mathbb{R}^3$ the system $[T](\mathbf{x}) = \mathbf{b}$ has a solution. From the discussion in Example 4 we know that this subspace consists of those vectors $\mathbf{b} = (b_1, b_2, b_3) \in \mathbb{R}^3$ whose coordinates satisfy

We can solve Condition (3) for b_2 and we get

$$b_2 = 2b_1 + 7b_3$$

so that $\operatorname{im} T$ has the parametric description

$$\mathbf{b} = s(1,2,0) + t(0,7,1), \quad s,t \in \mathbb{R}$$

In other words,

$$\operatorname{im} T = \langle (1, 2, 0), (0, 7, 1) \rangle$$

and since $\{(1, 2, 0), (0, 7, 1)\}$ is linearly independent, it is a basis of im T. So im T is 2-dimensional.

In the last example we found a basis for the range of T, by explicitly figuring under what conditions on **b** the system $[T]\mathbf{x} = \mathbf{b}$ has solutions. We could have avoided all that work by reasoning as follows: The columns of the matrix [T] are the (coordinates of the) images of the standard basis of \mathbb{R}^n , so im T is generated by (the vectors whose coordinates are) the columns of T. This means that the pivot columns of [T] form a basis of im T.¹ So a basis for the range of T in Example 4 is formed by the columns of [T] (since all columns are pivot). So a basis for im T is $\{(-2, 3, 1), (11, -13, 5)\}$.

In general we have:

Proposition 11. Let $T: \mathbb{R}^n \to \mathbb{R}^m$ be a linear transformation. Then The columns of [T] that correspond to the pivot columns of the RREF form a basis of the subspace im T.

A basis of ker T, is obtained by putting the solution of the homogeneous system $[T]\mathbf{x} = \mathbf{0}$ in to parametric form. There is one basic vector for every non-pivot column.

Definition 12. The *rank* of a linear transformation T is the dimension of its range:

$$r(T) = \dim \operatorname{im} T$$

The *nullity* of a linear transformation T is the dimension of its kernel:

$$n(T) = \dim \ker T$$

We immediately have:

Proposition 13. The rank of a linear transformation T equals the number of the pivot columns of [T]. The nullity of T equals the number of the non-pivot columns of [T].

If $T: \mathbb{R}^n \to \mathbb{R}^m$ then

$$r(T) + n(T) = n$$

 1 Why?

Example 14. Find the rank and nullity of the following 3×5 matrix:

$$\begin{pmatrix} 1 & 0 & -3 & 2 & 1 \\ 2 & 1 & 3 & -4 & 0 \\ 3 & 1 & 0 & -2 & 1 \end{pmatrix}$$

Answer. The RREF is

$$\begin{pmatrix} 1 & 0 & -3 & 2 & 1 \\ 0 & 1 & 9 & -8 & -2 \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$

There are two pivot and three non-pivot columns. Therefore r(T) = 2 and n(T) = 3.

Question 15. Find a basis for $\operatorname{im} T$ and $\ker T$ for the linear transformation of Example 14.

Finally, we give an example of a linear transformation that is both injective and surjective, in other words an isomorphism.

Example 16. The linear transformation $T: \mathbb{R}^3 \to \mathbb{R}^3$ whose matrix is given below is an isomorphism:

$$[T] = \begin{pmatrix} 1 & 2 & 3\\ 2 & 3 & -1\\ 3 & 0 & 2 \end{pmatrix}$$

Indeed the RREF is the 3×3 identity matrix. So every system $[T]\mathbf{x} = \mathbf{b}$ has a unique solution.

We will talk more about isomorphisms in the next lecture.

2. PROJECT: ORTHOGONAL COMPLEMENTS

Start with the first example from Section 1. Recall that the kernel of the transformation $T: \mathbb{R}^2 \to \mathbb{R}$ is the line K with equation x - y = 0. The line passing through the origin and orthogonal to K is the line with equation x + y = 0. Let's denote that line with K^{\perp} . Notice that K^{\perp} is a subspace of \mathbb{R}^2 .

Notice also that for any $b \in \mathbb{R}$ there are infinitely many vectors $\mathbf{x} \in \mathbb{R}^2$ with $T(\mathbf{x}) = b$, but *only one* of all these vectors lies in K^{\perp} . This allow us to chose a unique canonical solution to the equation x - y = b: just chose the vector that is perpendicular to the kernel.

Notice also that if we restrict T to have domain K^{\perp} , i.e. if we consider $T: K^{\perp} \to \mathbb{R}$ then T is an isomorphism.

We have an analogous situation for any linear transformation. In this project we develop this idea.

Definition 17. Let V be a linear subspace of \mathbb{R}^n . The orthogonal complement of V is defined as

$$V^{\perp} = \{ \mathbf{u} \in \mathbb{R}^n : \mathbf{u} \cdot \mathbf{v} = 0 \}$$



FIGURE 2. The transformation T of Example 2, again

Exercises.

- (1) Prove that V^{\perp} is a linear subspace of \mathbb{R}^n .
- (2) Prove that the only vector that is common to V and V[⊥] is the zero vector:

$$V \cap V^{\perp} = \{\mathbf{0}\}$$

(3) Prove that if $\{\mathbf{v}_1, \dots, \mathbf{v}_k\}$ is a basis of V then

$$V^{\perp} = \{ \mathbf{u} \in \mathbb{R}^n : \mathbf{u} \cdot \mathbf{v}_i = 0, \quad i = 1, \dots, k \}$$

(4) Prove that $\dim V^{\perp} = n - \dim V$

Hint. The previous question gives a system whose solution set is V^{\perp}

- (5) If $T: \mathbb{R} \to \mathbb{R}^m$ is a linear transformation, then if we restrict T to the orthogonal complement of the kernel, T becomes injective. In other words $T: (\ker T)^{\perp} \to \mathbb{R}^m$ is injective.
- (6) Find a basis for $(\ker T)^{\perp}$ where T is the linear transformation of Example 3.
- (7) This exercise also refers to Example 3. For $\mathbf{b} = (3, 4)$ find the unique $\mathbf{x} \in (\ker T)^{\perp}$ so that $T(\mathbf{x}) = \mathbf{b}$.