Linear Transformations

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1 Introduction

To introduce this topic we start with the idea of a function. The generic function f(x) = y contains an independent variable x and a dependent variable y. Assigning to x some number and using it in the function's calculation gives us a unique value y.

This concept extends to the realm of vectors. Instead of treating an individual number as an independent variable, we use an input vector. In this document we'll call this \vec{v} . A matrix will encode the calculation being performed. This is traditionally called matrix A. The result of this operation is matrix B. This is written in the form of a matrix multiplication:

$$A\vec{v} = B$$

We can also describe the environment in which this operation takes place. This is often written in the following form:

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

The letter T denotes that this is describing a transformation. The input vector is from the vector space \mathbb{R}^n . This space is known as the **domain**. If we have, for instance, \mathbb{R}^2 we are saying that this transformation accepts a two dimensional vector with a real x and y component. Next we define what the transformation itself does.

Here is an example, where $T: \mathbb{R}^2 \to \mathbb{R}^2$

$$T\begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 2x \\ 3y \end{pmatrix}$$

In this example, feeding the transformation a two dimension vector, say < 1, 3 > results in another two dimension vector < 2, 9 >.

It's also possible to rewrite the transformation in its standard form $A\vec{v} = B$. To do this we will use the standard coordinate vectors:

$$e_1 = \begin{pmatrix} 1 \\ 0 \\ 0 \\ \dots \\ 0 \end{pmatrix}, e_2 = \begin{pmatrix} 0 \\ 1 \\ 0 \\ \dots \\ 0 \end{pmatrix}, e_3 = \begin{pmatrix} 0 \\ 0 \\ 1 \\ \dots \\ 0 \end{pmatrix}$$

The number of vectors needed is the same as the dimensions of the input vector. The matrix A, which encapsulates the transformation is composed of the columns that result from $T(e_n)$.

We can now rewrite the following transformation:

$$T \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 2x \\ 3y \end{pmatrix}$$
$$T(e_1) = T \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \begin{pmatrix} 2 \\ 0 \end{pmatrix}, T(e_2) = T \begin{pmatrix} 0 \\ 1 \end{pmatrix} = \begin{pmatrix} 0 \\ 3 \end{pmatrix}$$

So we get:

$$A = \begin{pmatrix} 2 & 0 \\ 0 & 3 \end{pmatrix}$$

We can now write the transformation in its standard form:

$$A = \begin{pmatrix} 2 & 0 \\ 0 & 3 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = B$$

2 Image and Kernel

The image is the vector space that the transformation will end up occupying. If the definition of the transformation is given, this is a trivial process. In the example we've been using:

$$T\begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 2x \\ 3y \end{pmatrix}$$

We can describe the **image** as: x = 2s and y = 2t, where $s, t \in \mathbb{R}$. Now suppose you were not given the transformation definition but only matrix A. We can arrive at the transformation by simply carrying out the matrix multiplication of the standard form:

$$A = \begin{pmatrix} 2 & 0 \\ 0 & 3 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 2x \\ 3y \end{pmatrix}$$

We can describe the **kernel** as the solution to $A\vec{v} = \vec{0}$.

For the transformation we've been looking at, the only way to obtain the zero vector is if x = 0 and y = 0. We can therefore say:

$$Ker(T) = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

For more complex transformations, applying row elimination can be used to formulate the kernel. Consider the following transformation:

$$T: \mathbb{R}^3 \to \mathbb{R}^2$$

$$T\begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} x - z \\ x + 2y - 4z \end{pmatrix}$$

To determine the kernel we solve for the system of equations comprised of x - z = 0 and x + 2y - 4z = 0. Row reduction can be used here:

$$\begin{pmatrix} 1 & 0 & -1 \\ 1 & 2 & -4 \end{pmatrix} \xrightarrow{R_2 - R_1 = R_2} \begin{pmatrix} 1 & 0 & -1 \\ 0 & 2 & -3 \end{pmatrix} \xrightarrow{\frac{1}{2}R_2 = R_2} \begin{pmatrix} 1 & 0 & -1 \\ 0 & 1 & -\frac{3}{2} \end{pmatrix}$$
$$Ker(T) = \left\{ \begin{array}{c} x = s \\ y = \frac{3}{2}s \\ z = s \end{array} \right\}, s \in \mathbb{R}$$

The key point is that for any value of s the x, y, z values produced, when plugged into the transformation will always result in the zero vector.

3 Set Visualization

We can now draw a diagram that brings together all these terms:

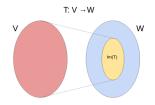


Figure 1

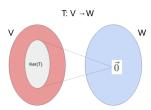


Figure 2

In figure 1, V represents the domain and W the codomain. The possible set of vectors produced by the transformation are represented by the yellow subset within W. This subset is the image of T. In figure 2, there is a subset within the domain that will always be transformed into the zero vector, thus the gray subset within V is the kernel of T.

4 Linear Transformation

Although we can use this terminology to describe all transformations, not all transformations are linear. A linear transformation must meet certain conditions.

 $T: \mathbb{R}^n \to \mathbb{R}^m$ is a linear transformation if the following conditions hold:

- $T(\vec{u} + \vec{v}) = T(\vec{u}) + T(\vec{v})$, where $\vec{u}, \vec{v} \in \mathbb{R}^m$
- $T(k\vec{v}) = kT(\vec{v})$ for a scalar k.

Let's demonstrate that the two transformations we've examined are indeed linear transformations. Let $\vec{u} = \langle a, b \rangle$ and $\vec{v} = \langle c, d \rangle$ and k a scalar:

$$T \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} 2x \\ 3y \end{pmatrix}$$
$$T(\vec{u}) + T(\vec{v}) = \begin{pmatrix} 2a \\ 3b \end{pmatrix} + \begin{pmatrix} 2c \\ 3d \end{pmatrix} = \begin{pmatrix} 2(a+c) \\ 3(b+d) \end{pmatrix}$$
$$T(\vec{u} + \vec{v}) = \begin{pmatrix} 2(a+c) \\ 3(b+d) \end{pmatrix}$$

First condition met

$$T(k\vec{u}) = k \begin{pmatrix} 2a \\ 3b \end{pmatrix} = T(k\vec{u}) = \begin{pmatrix} 2ak \\ 3bk \end{pmatrix}$$
$$kT(\vec{u}) = k \begin{pmatrix} 2a \\ 3b \end{pmatrix} = \begin{pmatrix} 2ak \\ 3bk \end{pmatrix}$$

Second condition met

$$T \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} x - z \\ x + 2y - 4z \end{pmatrix}$$

$$T(\vec{u}) + T(\vec{v}) = \begin{pmatrix} a - c \\ a + 2b - 4c \end{pmatrix} + \begin{pmatrix} d - f \\ d + 2e - 4f \end{pmatrix}$$

$$= \begin{pmatrix} a - c + d - f \\ a + 2b - 4c + d + 2d - 4f \end{pmatrix}$$

$$T(\vec{u} + \vec{v}) = \begin{pmatrix} (a + d) - (c + f) \\ (a + d) + 2(b - e) - 4(c + f) \end{pmatrix}$$

$$= \begin{pmatrix} a - c + d - f \\ a + 2b - 4c + d + 2d - 4f \end{pmatrix}$$

First condition met

$$T(k\vec{u}) = \begin{pmatrix} ak - ck \\ ak + 2yk - 4zk \end{pmatrix} = k \begin{pmatrix} a - c \\ a + 2y - 4z \end{pmatrix}$$
$$kT(\vec{u}) = k \begin{pmatrix} a - c \\ a + 2y - 4z \end{pmatrix}$$

Second condition met

5 Additional Problems

Ex. 1 Is the following a linear transformation?

$$T: \mathbb{R}^2 \to \mathbb{R}^2$$

$$T\begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} x^2 \\ y^2 \end{pmatrix}$$

Let $\vec{u} = \langle a, b \rangle$ and $\vec{v} = \langle c, d \rangle$.

We test the first condition:

$$T(\vec{u}) + T(\vec{v}) = \begin{pmatrix} a^2 + c^2 \\ b^2 + d^2 \end{pmatrix}$$

$$T(\vec{u} + \vec{v}) = \begin{pmatrix} (a+c)^2 \\ (b+d)^2 \end{pmatrix} = \begin{pmatrix} a^2 + 2ac + c^2 \\ b^2 + 2bd + d^2 \end{pmatrix}$$

Since $T(\vec{u}) + T(\vec{v}) \neq T(\vec{u} + \vec{v})$, this is not a linear transformation

Ex. 2 Is the following a linear transformation?

$$T: \mathbb{R}^3 \to \mathbb{R}^4$$

$$T\begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} x+y \\ y+z \\ x-z \\ z-y \end{pmatrix}$$

Let $\vec{u} = \langle a, b \rangle$ and $\vec{v} = \langle c, d \rangle$.

We test the first condition:

$$T(\vec{u}) + T(\vec{v}) = \begin{pmatrix} a+b+d+e \\ b+c+e+f \\ a-c+d-f \\ c-b+f-e \end{pmatrix}$$

$$T(\vec{u} + \vec{v}) = \begin{pmatrix} a+d+b+e \\ b+e+c+f \\ (a+d)-(c+f) \\ (c+f)-(b+e) \end{pmatrix} = \begin{pmatrix} a+b+d+e \\ b+c+e+f \\ a-c+d-f \\ c-b+f-e \end{pmatrix}$$

Condition Met

We test the second condition:

$$T(k\vec{u}) = \begin{pmatrix} ak + bk \\ bk + ck \\ ak - ck \\ ck - bk \end{pmatrix} = k \begin{pmatrix} a + b \\ b + c \\ a - c \\ c - b \end{pmatrix}$$

$$kT(\vec{u}) = k \begin{pmatrix} a+b \\ b+c \\ a-c \\ c-b \end{pmatrix}$$

Condition Met

T is a linear transformation.

Ex. 3 For T in example 2, determine the kernel.

We start by reducing the transformation using a matrix:

$$T \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} x+y \\ y+z \\ x-z \\ z-y \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \\ 1 & 0 & -1 \\ 0 & -1 & 1 \end{pmatrix}$$

$$\begin{pmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \\ 1 & 0 & -1 \\ 0 & -1 & 1 \end{pmatrix} \xrightarrow{R_1 - R_2 = R_1} \begin{pmatrix} 1 & 0 & -1 \\ 0 & 1 & 1 \\ 1 & 0 & -1 \\ 0 & -1 & 1 \end{pmatrix} \xrightarrow{R_3 - R_1 = R_3} \begin{pmatrix} 1 & 0 & -1 \\ 0 & 1 & 1 \\ 0 & 0 & 0 \\ 0 & -1 & 1 \end{pmatrix} \xrightarrow{R_2 + R_4 = R_2}$$

$$\begin{pmatrix} 1 & 0 & -1 \\ 0 & 1 & 1 \\ 0 & 0 & 0 \\ 0 & -1 & 1 \end{pmatrix} \xrightarrow{\frac{1}{2}R_2 = R_2} \begin{pmatrix} 1 & 0 & -1 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \\ 0 & -1 & 1 \end{pmatrix} \xrightarrow{R_1 + R_2 = R_1} \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \\ 0 & -1 & 1 \end{pmatrix} \xrightarrow{R_4 - R_2 = R_4} \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix}$$

So

$$Ker(T) = \left\{ \begin{array}{l} x = 0 \\ y = 0 \\ z = 0 \end{array} \right\}$$

Ex. 4 Given the following matrix A, find the kernel:

$$A = \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{pmatrix}$$
$$3 \setminus \begin{pmatrix} x \\ x \end{pmatrix} \quad \begin{pmatrix} x + 2y + 4y \\ x \end{pmatrix}$$

$$\begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{pmatrix} \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} x + 2y + 3z \\ 4x + 5y + 6z \\ 7x + 8y + 9z \end{pmatrix}$$

The kernel is the solution to the following system:

$$x + 2y + 3z = 0$$
$$4x + 5y + 6z = 0$$
$$7x + 8y + 9z = 0$$

... which can be achieved through row reduction:

$$\begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{pmatrix} \xrightarrow{R_2 - 4R_1 = R_2} \begin{pmatrix} 1 & 2 & 3 \\ 0 & -3 & -6 \\ 7 & 8 & 9 \end{pmatrix} \xrightarrow{-\frac{1}{3}R_2 = R_2} \begin{pmatrix} 1 & 2 & 3 \\ 0 & 1 & 2 \\ 7 & 8 & 9 \end{pmatrix} \xrightarrow{R_3 - 7R_1 = R_3}$$

$$\begin{pmatrix} 1 & 2 & 3 \\ 0 & 1 & 2 \\ 0 & -6 & -12 \end{pmatrix} \xrightarrow{R_3 + 6R_2 = R_3} \begin{pmatrix} 1 & 2 & 3 \\ 0 & 1 & 2 \\ 0 & 0 & 0 \end{pmatrix} \xrightarrow{R_1 - 2R_2 = R_1} \begin{pmatrix} 1 & 0 & -1 \\ 0 & 1 & 2 \\ 0 & 0 & 0 \end{pmatrix}$$

We are left with a system of equations that contains x - z = 0 and y + 2z = 0, so the solution is:

$$Ker(T) = \left\{ \begin{array}{c} x = s \\ y = -2s \end{array} \right\}, s \in \mathbb{R}$$

6 Image Transformation

An application of matrix transformation is found in computer graphics. Image editing tools can scale, shear, and rotate a particular image. In this section, we will cover examples of scaling and shearing. Suppose we have the image below, and every pixel is represented by a vector.



Figure 3

6.1 Scaling

In the following example, we describe a 1.5x vertical scaling using $A\vec{v} = B$:

$$\begin{pmatrix} 1 & 0 \\ 0 & \frac{3}{2} \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = B$$

For $\vec{v} = <1, 1>$, this transformation yields <1, 1.5>. On the diagram below, we see on the left the result of two of these vectors in addition to the vector $\vec{v} = <-1, -1>$. On the right, we see the same representation but only showing the tips of the vectors.

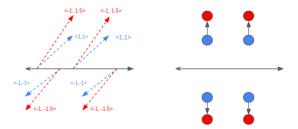


Figure 4

If we can imagine every pixel of the image as being the tip of a vector, and apply the transformation described above we will see that the image stretches vertically:

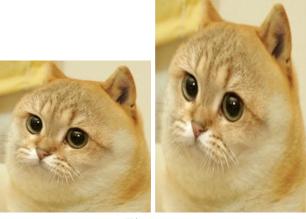


Figure 5

6.2 Shearing

We will now consider the following transformation:

$$\begin{pmatrix} 1 & 1.5 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = B$$

Supposed we provided $\vec{v}=<1,1>$ to the transformation, in return we would get <2.5,1>. The diagram below shows the result of transforming a few vectors:

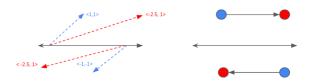


Figure 6

If we apply this shear to all the pixels on an image, we would observe the following:



Figure 7