

# LINEAR ALGEBRA: LECTURE 22–24

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## 1. A COMMENT ON NOTATION

Thus far, we've been writing points in the form  $(a, b)$  or  $\begin{pmatrix} a \\ b \end{pmatrix}$  and vectors in the form  $a\vec{e}_1 + b\vec{e}_2$ . I've been insisting on this to emphasize the distinction between points and vectors. However, as we've seen, linear maps can't distinguish between the two. Since this is a course on linear algebra (and therefore concerned almost exclusively with linear maps) we will use point notation to denote both points and vectors. In other words, we will write  $\begin{pmatrix} a \\ b \end{pmatrix}$  instead of  $a\vec{e}_1 + b\vec{e}_2$ .

## 2. AN EXPLICIT FORMULA FOR FIBONACCI NUMBERS

Recall that we are trying to find a formula for the  $n^{\text{th}}$  Fibonacci number  $f_n$  using matrices. Our strategy (described last time) is to find a diagonal matrix which is similar to  $\begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix}$ , say,

$$\begin{pmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{pmatrix} = P^{-1} \begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix} P. \quad (1)$$

We thus rephrase our goal as follows:

**Question 1.** *Do there exist  $\lambda_1, \lambda_2 \in \mathbb{R}$  and an invertible matrix  $P$  such that (1) holds?*

If the answer to this question is affirmative, then it would follow that

$$\begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix} P = P \begin{pmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{pmatrix}$$

whence

$$\begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix} P(\vec{e}_1) = P \begin{pmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{pmatrix} (\vec{e}_1) = P(\lambda_1 \vec{e}_1) = \lambda_1 P(\vec{e}_1). \quad (2)$$

Further, note that  $P(\vec{e}_1) \neq \mathbf{0}$ , since we are hoping to find an invertible map  $P$ . (Can you justify this sentence?) Thus, to have any hope of answering Question 1 in the affirmative, we must be able to find a number  $\lambda$  and a vector  $\vec{v}$  such that

$$\begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix} \vec{v} = \lambda \vec{v}. \quad (3)$$

In other words, we wish to find some nonzero vector  $\vec{v}$  such that when we apply  $\begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix}$  to it we get the same vector back, just stretched out by a factor of  $\lambda$ . Note that no matter what  $\lambda$  and  $\vec{v}$  are, there's one map which has the desired effect:

$$\begin{pmatrix} \lambda & 0 \\ 0 & \lambda \end{pmatrix} \vec{v} = \lambda \vec{v}.$$

Thus, we wish to find  $\lambda$  and  $\vec{v}$  such that

$$\begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix} \vec{v} = \begin{pmatrix} \lambda & 0 \\ 0 & \lambda \end{pmatrix} \vec{v}$$

or equivalently

$$\begin{pmatrix} 1 - \lambda & 1 \\ 1 & -\lambda \end{pmatrix} \vec{v} = \mathbf{0}.$$

Recall from above that  $\vec{v} \neq \mathbf{0}$ . What sort of linear map sends a nonzero vector to the zero vector? Only a singular one! (Can you explain why?) Thus, we deduce that we must have

$$\det \begin{pmatrix} 1 - \lambda & 1 \\ 1 & -\lambda \end{pmatrix} = 0,$$

or in other words,

$$\lambda^2 - \lambda - 1 = 0.$$

Recall that we're searching for a number  $\lambda$  and a nonzero vector  $\vec{v}$  which satisfy the relation (3). What we've just proved is that if these exist, then

$$\lambda = \frac{1 \pm \sqrt{5}}{2}$$

Set

$$\lambda_1 := \frac{1 + \sqrt{5}}{2} \quad \text{and} \quad \lambda_2 := \frac{1 - \sqrt{5}}{2}$$

Returning to (3), it remains to find some nonzero vector  $\vec{v}_1$  such that

$$\begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix} \vec{v}_1 = \lambda_1 \vec{v}_1.$$

How do we construct such a vector? The most natural approach is to write  $\vec{v}_1 = \begin{pmatrix} x \\ y \end{pmatrix}$ , plug it in above, and solve for  $x$  and  $y$ . When we did this in class, we discovered that  $x = \lambda_1 y$  and that  $y$  satisfies  $(\lambda_1^2 - \lambda_1 - 1)y = 0$ . Note that the latter relationship holds for every  $y$ ! (Why is this?) Although this looks like a failure at first glance, it's actually a success – this tells us that we can choose  $y$  to be anything, and then set  $x = \lambda_1 y$ . For example, we can take

$$\vec{v}_1 := \begin{pmatrix} \lambda_1 \\ 1 \end{pmatrix}$$

With this choice, it is easy to verify that

$$\begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix} \vec{v}_1 = \lambda_1 \vec{v}_1.$$

Going back to (2) shows that we'd like to find an invertible map  $P$  such that

$$P(\vec{e}_1) = \vec{v}_1 = \begin{pmatrix} \lambda_1 \\ 1 \end{pmatrix}$$

The exact same arguments show that we'd like

$$P(\vec{e}_2) = \begin{pmatrix} \lambda_2 \\ 1 \end{pmatrix}$$

Combining the two previous statements tells us how to choose  $P$ :

$$P = \begin{pmatrix} \lambda_1 & \lambda_2 \\ 1 & 1 \end{pmatrix}$$

Now that we've determined  $\lambda_1$ ,  $\lambda_2$ , and  $P$ , it's straightforward to verify (1).

Having done all of this, it's not so hard to find an explicit formula for the  $n^{\text{th}}$  Fibonacci number. Manipulating (1), we see that

$$\begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix} = P \begin{pmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{pmatrix} P^{-1}$$

(make sure you can explain why!). Raising both sides to the  $n^{\text{th}}$  power gives

$$\begin{pmatrix} f_{n+1} & f_n \\ f_n & f_{n-1} \end{pmatrix} = \begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix}^n = P \begin{pmatrix} \lambda_1^n & 0 \\ 0 & \lambda_2^n \end{pmatrix} P^{-1} = \begin{pmatrix} \lambda_1 & \lambda_2 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} \lambda_1^n & 0 \\ 0 & \lambda_2^n \end{pmatrix} \begin{pmatrix} \lambda_1 & \lambda_2 \\ 1 & 1 \end{pmatrix}^{-1} = \begin{pmatrix} * & * \\ \frac{\lambda_1^n - \lambda_2^n}{\lambda_1 - \lambda_2} & * \end{pmatrix}$$

It follows that

$$f_n = \frac{\lambda_1^n - \lambda_2^n}{\lambda_1 - \lambda_2}$$

where  $\lambda_1, \lambda_2$  are the two solutions to the equation  $\lambda^2 - \lambda - 1 = 0$ . We've discovered our formula!

### 3. SPECTRAL THEORY

The key to figuring out the formula for  $f_n$  above was finding numbers  $\lambda_1$  and  $\lambda_2$ , along with an invertible matrix  $P$ , such that

$$\begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix} = P \begin{pmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{pmatrix} P^{-1}$$

In addition to being useful for calculating powers of  $\begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix}$ , this gives us a nice geometric interpretation of the action of  $\begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix}$  on the plane. Recall from above that

$$\begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix} \vec{v}_j = \lambda_j \vec{v}_j$$

where  $P(\vec{e}_1) = \vec{v}_1$  and  $P(\vec{e}_2) = \vec{v}_2$ . In other words,  $P$  is the change-of-basis map from  $\vec{e}_1, \vec{e}_2$  to  $\vec{v}_1, \vec{v}_2$ , and if we replace the usual coordinate system (generated by  $\vec{e}_1$  and  $\vec{e}_2$ ) by the one generated by  $\vec{v}_1$  and  $\vec{v}_2$ , then  $\begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix}$  has a simple geometric description: it stretches the plane out by a factor of  $\lambda_1$  in the  $\vec{v}_1$  direction and by  $\lambda_2$  in the  $\vec{v}_2$  direction.

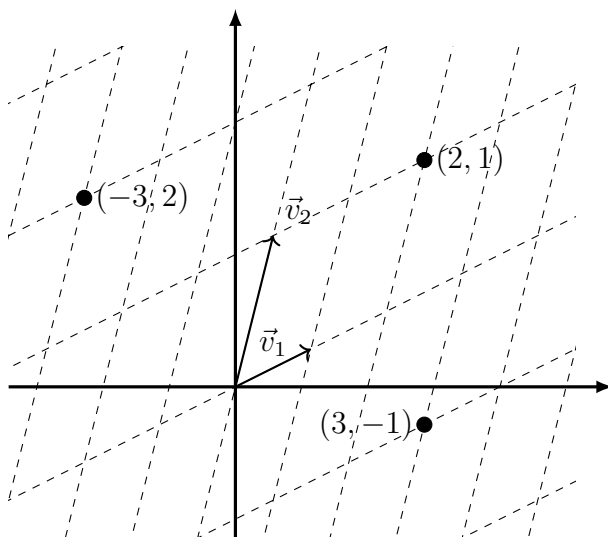
Let's generalize this. Suppose  $f : \mathbb{R}^2 \rightarrow \mathbb{R}^2$  is a linear map. Suppose we can find numbers  $\lambda_1, \lambda_2$  and an invertible matrix  $P$  such that

$$f = P \begin{pmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{pmatrix} P^{-1}.$$

This is called the *spectral decomposition* of  $f$ , and it gives us a nice way to interpret  $f$ . Think of  $P$  as the change-of-basis matrix from  $\vec{e}_1, \vec{e}_2$  to some vectors  $\vec{v}_1, \vec{v}_2$ . Then it is straightforward to verify that

$$f(\vec{v}_1) = \lambda_1 \vec{v}_1 \quad \text{and} \quad f(\vec{v}_2) = \lambda_2 \vec{v}_2$$

Now label each point of the plane in terms of how to get there using  $\vec{v}_1$  and  $\vec{v}_2$ . For example:



Once we adopt this perspective, it's very easy to describe what  $f$  does to any point: it stretches the first coordinate by  $\lambda_1$  and the second by  $\lambda_2$ . For example, where does  $f$  send the point  $(3, -2)$  indicated above? (Note: this isn't the usual  $(3, -2)$ ; it's  $3\vec{v}_1 - 2\vec{v}_2$ .) Easy:  $f(3, -2) = (3\lambda_1, -2\lambda_2)$ . The quantities  $\lambda_j$  and  $\vec{v}_j$  play a pivotal role in understanding the spectral decomposition, so they get a special name.

**Definition.** Given a linear map  $f : \mathbb{R}^2 \rightarrow \mathbb{R}^2$ . We say a number  $\lambda$  is an *eigenvalue* of  $f$  if and only if there exists a nonzero vector  $\vec{v}$  such that

$$f(\vec{v}) = \lambda\vec{v}.$$

In this case, we say  $\vec{v}$  is an *eigenvector* corresponding to the eigenvalue  $\lambda$ .

**Example 1.** We discovered above that the map  $\begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix}$  has eigenvalues

$$\lambda_1 = \frac{1 + \sqrt{5}}{2} \quad \text{and} \quad \lambda_2 = \frac{1 - \sqrt{5}}{2}$$

and that their corresponding eigenvectors are

$$\vec{v}_1 = \begin{pmatrix} \lambda_1 \\ 1 \end{pmatrix} \quad \text{and} \quad \vec{v}_2 = \begin{pmatrix} \lambda_2 \\ 1 \end{pmatrix}$$

One immediate remark is that eigenvectors aren't uniquely determined.

**Proposition 1.** Suppose  $f$  has eigenvalue  $\lambda$  with corresponding eigenvector  $\vec{v}$ . Then  $\alpha\vec{v}$  is an eigenvector corresponding to  $\lambda$  for every  $\alpha \neq 0$ .

**3.1. Finding the spectral decomposition.** We now generalize the process used to analyze  $\begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix}$  to determine the spectral decomposition of an arbitrary linear map  $f : \mathbb{R}^2 \rightarrow \mathbb{R}^2$ . We break the process into a few steps.

STEP 1. Solve the equation  $\det(f - \lambda I) = 0$  for  $\lambda$ , where  $I = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$  is the identity matrix. The eigenvalues of  $f$  are the solutions to this equation.

*Why does this work? Suppose  $\lambda$  is an eigenvalue of  $f$ . Then by definition, there exists a nonzero  $\vec{v}$  such that  $f(\vec{v}) = \lambda\vec{v}$ . Just as in the Fibonacci example, this happens iff  $\exists \vec{v} \neq \mathbf{0}$  such that  $(f - \lambda I)\vec{v} = \mathbf{0}$ . But this occurs iff the map  $f - \lambda I$  is singular.*

STEP 2. Solve the equation  $f(x, 1) = (\lambda x, \lambda)$  for  $x$ . Then  $\vec{v} := \begin{pmatrix} x \\ 1 \end{pmatrix}$  is an eigenvector corresponding to  $\lambda$ .

*Why does this work? We found an eigenvalue  $\lambda$  above, and we wish to find a corresponding eigenvector  $\vec{v}$ . Since any rescaling of  $\vec{v}$  remains an eigenvector, we may as well rescale in such a way that  $\vec{v} = \begin{pmatrix} x \\ 1 \end{pmatrix}$ . Now by definition,  $\vec{v}$  must satisfy the equation  $f(\vec{v}) = \lambda\vec{v}$ .*

STEP 3. Say the two eigenvalues of  $f$  are  $\lambda_1$  and  $\lambda_2$ , with corresponding eigenvectors  $\vec{v}_1 = \begin{pmatrix} a \\ c \end{pmatrix}$  and  $\vec{v}_2 = \begin{pmatrix} b \\ d \end{pmatrix}$ .

Let  $P := \begin{pmatrix} a & b \\ c & d \end{pmatrix}$ . Then  $f = P \begin{pmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{pmatrix} P^{-1}$ .

*Why is this? A good exercise!*

Before exploring specific examples, let's try to predict what sorts of problems might arise in the steps above. In Step 1, we might only find a single eigenvalue; when this happens, this usually bodes ill for the spectral decomposition, as we shall see below. In Step 2, it's possible that  $\vec{e}_1$  is an eigenvector, in which case no renormalization would make it possess the form we described. Finally, in Step 3, we need to worry about the

possibility that  $P$  is not invertible. These fears are all justified, and some of these problems are fatal to the process. In fact, as we shall see, *not every linear map admits a spectral decomposition*. By contrast, every linear map admits a Singular Value Decomposition. (On the other hand, when a map *does* admit a spectral decomposition, it's much easier to find than the SVD.)

Let's explore a few representative examples.

EX. 1.  $f = \begin{pmatrix} 1 & 3 \\ 5 & 3 \end{pmatrix}$

To find the eigenvalues, we first solve the equation

$$\det(f - \lambda I) = 0$$

for  $\lambda$ . The LHS is

$$\det \left( \begin{pmatrix} 1 & 3 \\ 5 & 3 \end{pmatrix} - \begin{pmatrix} \lambda & 0 \\ 0 & \lambda \end{pmatrix} \right) = \det \begin{pmatrix} 1-\lambda & 3 \\ 5 & 3-\lambda \end{pmatrix} = (1-\lambda)(3-\lambda) - 15.$$

Expanding this, setting equal to zero, and solving yields  $\lambda = -2, 6$ . Let's set

$$\lambda_1 := -2 \quad \text{and} \quad \lambda_2 := 6.$$

These are the eigenvalues.

Next, we find corresponding eigenvectors. We first solve the equation

$$\begin{pmatrix} 1 & 3 \\ 5 & 3 \end{pmatrix} \begin{pmatrix} x \\ 1 \end{pmatrix} = -2 \begin{pmatrix} x \\ 1 \end{pmatrix}$$

From this we easily deduce that  $x = -1$ , whence our first eigenvector is  $\vec{v}_1 := \begin{pmatrix} -1 \\ 1 \end{pmatrix}$ .

Similarly, solving

$$\begin{pmatrix} 1 & 3 \\ 5 & 3 \end{pmatrix} \begin{pmatrix} x \\ 1 \end{pmatrix} = 6 \begin{pmatrix} x \\ 1 \end{pmatrix}$$

yields  $x = 3/5$ , whence  $\vec{v}_2 := \begin{pmatrix} 3/5 \\ 1 \end{pmatrix}$ . If we wish, we can make this look nicer by rescaling it to  $\vec{v}_2 := \begin{pmatrix} 3 \\ 5 \end{pmatrix}$ .

The final step of the process is to determine the change of basis map  $P$ :

$$P := \begin{pmatrix} -1 & 3 \\ 1 & 5 \end{pmatrix}$$

Thus, our spectral decomposition is

$$\begin{pmatrix} 1 & 3 \\ 5 & 3 \end{pmatrix} = P \begin{pmatrix} -2 & 0 \\ 0 & 6 \end{pmatrix} P^{-1}$$

EX. 2.  $R_{\pi/2} = \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix}$

As before, we begin by finding the eigenvalues of  $R_{\pi/2}$  via the equation

$$\det(R_{\pi/2} - \lambda I) = 0.$$

This equation can be rewritten as

$$\lambda^2 + 1 = 0,$$

so the eigenvalues of  $R_{\pi/2}$  are  $\lambda_1 = i$  and  $\lambda_2 = -i$ . Next, we find the corresponding eigenvectors. Write  $\vec{v}_1 = \begin{pmatrix} x \\ 1 \end{pmatrix}$ ; we're supposed to solve

$$\begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} x \\ 1 \end{pmatrix} = i \begin{pmatrix} x \\ 1 \end{pmatrix},$$

which immediately yields  $x = i$ . It follows that  $\vec{v}_1 = \begin{pmatrix} i \\ 1 \end{pmatrix}$ . A similar argument shows that  $\vec{v}_2 = \begin{pmatrix} -i \\ 1 \end{pmatrix}$ . Finally, let  $P$  be the change-of-basis

$$P = \begin{pmatrix} i & -i \\ 1 & 1 \end{pmatrix}.$$

Then we have the spectral decomposition

$$R_{\pi/2} = P \begin{pmatrix} i & 0 \\ 0 & -i \end{pmatrix} P^{-1}.$$

Note that we've successfully found a spectral decomposition of  $R_{\pi/2}$ , but only if we allow ourselves to use imaginary numbers. This is a bit odd, since the original function  $R_{\pi/2}$  has nothing to do with imaginary numbers! This hints at a connection between rotations in  $\mathbb{R}^2$  and complex numbers. On the other hand, a bit more thought shows that it's not unreasonable that the spectral decomposition of a rotation should be unusual, since a rotation doesn't stretch the plane in any direction.

EX. 3.  $g = \begin{pmatrix} 1 & 1 \\ -1 & 3 \end{pmatrix}$

Following the above procedure, we find that the only eigenvalue of  $g$  is  $\lambda = 2$ . Continuing along shows that the only eigenvectors of  $g$  are scalar multiples of  $\begin{pmatrix} 1 \\ 1 \end{pmatrix}$ . It follows that any change-of-basis matrix  $P$  would not be invertible, which means that  $g$  has no spectral decomposition. An alternative way to express this is that  $g$  is *not diagonalizable*.

Note that we could have seen that  $g$  wasn't diagonalizable without solving for the eigenvalues. For, suppose  $g$  did have a spectral decomposition. Since we know the only eigenvalue is 2, we would be able to write

$$g = P \begin{pmatrix} 2 & 0 \\ 0 & 2 \end{pmatrix} P^{-1}$$

for some matrix  $P$ . But this would immediately imply  $g = \begin{pmatrix} 2 & 0 \\ 0 & 2 \end{pmatrix}$ , which isn't the case! Put differently, we've just shown that the only diagonalizable matrix with both eigenvalues equal to 2 is  $2I$ .