

# Math 24: Winter 2021

## Lecture 18

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# Let's Get Started

- 1 We should be recording.
- 2 Remember, it is more comfortable for me if you turn on your video so that I feel like I am talking to real people.
- 3 Our Midterm will be available Thursday after office hours and must be turned in by Saturday, February 20th, at 10pm. No exceptions. It will cover through §4.3 in the text.
- 4 This time you will have 3.5 hours with 30 minutes uploading time. You should plan ahead and block out a time to take the exam now.
- 5 But first, are there any questions from last time?

# Diagonalizable Transformations

## Definition

A linear operator  $T$  on a finite-dimensional vector space  $V$  is said to be **diagonalizable** if there is an ordered basis  $\beta$  for  $V$  such that  $[T]_{\beta}$  is a diagonal matrix.

## Remark

Suppose that  $\beta = \{v_1, \dots, v_n\}$  is an ordered basis such that  $[T]_\beta = [[T(v_1)]_\beta \cdots [T(v_n)]_\beta]$  is diagonal. That is,

$$[[T(v_1)]_\beta \cdots [T(v_n)]_\beta] = \begin{pmatrix} \lambda_1 & 0 & \cdots & 0 \\ 0 & \lambda_2 & \cdots & 0 \\ \vdots & 0 & \ddots & 0 \\ 0 & 0 & \cdots & \lambda_n \end{pmatrix}$$

Then for each  $1 \leq k \leq n$ ,  $[T(v_k)]_\beta = \lambda_k e_k$  where  $\{e_1, \dots, e_n\}$  is the standard ordered basis for  $\mathbf{F}^n$ . That means  $T(v_k) = \lambda_k v_k$  for all  $k$ .

Conversely, if  $\beta = \{v_1, \dots, v_n\}$  is an ordered basis for  $V$  such that  $T(v_k) = \lambda_k v_k$  for all  $k$ , then  $[T]_\beta$  is diagonal just as above.

### Definition

Suppose that  $T$  is a linear operator on a vector space  $V$ . Then a **nonzero** vector  $v \in V$  is called an **eigenvector** for  $T$  if there is a scalar  $\lambda \in \mathbf{F}$  such that  $T(v) = \lambda v$ . We call  $\lambda$  the **eigenvalue** corresponding to the eigenvector  $v$ .

If  $A \in M_{n \times n}(\mathbf{F})$ , then we call a nonzero  $v \in \mathbf{F}^n$  an eigenvector for  $A$  if it is an eigenvector for  $L_A$  so that  $Av = \lambda v$  for some scalar  $\lambda$ . Again, we call  $\lambda$  the eigenvalue corresponding to the eigenvector  $v$ .

# Our First Eigen Theorem

We can summarize the previous discussion in a useful theorem.

## Theorem

*A linear operator  $T$  on a finite-dimensional vector space  $V$  is diagonalizable if and only if there is an ordered basis for  $V$  consisting of eigenvectors for  $T$ . If  $\beta = \{v_1, \dots, v_n\}$  is an ordered basis of eigenvectors for  $T$ , then  $[T]_\beta$  is the diagonal matrix  $D = (D_{ij})$  where*

$$D_{ij} = \begin{cases} \lambda_i & \text{if } i = j, \text{ and} \\ 0 & \text{if } i \neq j \end{cases}$$

*where  $\lambda_i$  is the eigenvalue corresponding to the eigenvector  $v_i$ .*

## Corollary

*A matrix  $A \in M_{n \times n}(\mathbf{F})$  is diagonalizable if and only if there is an ordered basis of  $\mathbf{F}^n$  consisting of eigenvectors for  $A$ . If  $\beta = \{v_1, \dots, v_n\}$  is an ordered basis of eigenvectors for  $A$  and if  $Q = [v_1 \ \cdots \ v_n]$  is the  $n \times n$ -matrix whose  $j^{\text{th}}$ -column is  $v_j$ , then  $D = Q^{-1}AQ$  is a diagonal matrix such that  $D_{ii}$  is the eigenvalue corresponding to  $v_i$ . In particular,  $A$  is diagonalizable if and only if it is similar to a diagonal matrix.*

## Proof.

The first statement follows from the previous theorem applied to  $T = L_A$ . If  $\beta = \{v_1, \dots, v_n\}$  is a basis of eigenvectors such that  $Av_i = D_{ii}v_i$ , then  $[L_A]_{\beta} = D$  as in the statement of the corollary. But  $Q = [I]_{\beta}^{\sigma}$  where  $\sigma$  is the standard ordered basis, and

$$D = [L_A]_{\beta} = [I]_{\sigma}^{\beta} [L_A]_{\sigma}^{\sigma} [I]_{\beta}^{\sigma} = Q^{-1}AQ.$$

## Proof Continued.

Conversely, if  $Q$  is invertible and  $D = Q^{-1}AQ$  is diagonal, then the columns of  $Q$ , say  $\{v_1, \dots, v_n\}$ , are an ordered basis for  $\mathbf{F}^n$  and  $Q = [I]_{\beta}^{\sigma}$ . Furthermore,

$$[L_A]_{\beta}^{\beta} = [I]_{\sigma}^{\beta} [L_A]_{\sigma}^{\sigma} [I]_{\beta}^{\sigma} = Q^{-1}AQ = D.$$

Thus  $L_A(v_i) = Av_i = D_{ii}v_i$  and  $\beta$  is a basis of eigenvectors for  $A$ . □



# Examples of Eigenvectors

## Example

Let  $A = \begin{pmatrix} 7 & -10 \\ 5 & -8 \end{pmatrix}$ ,  $v_1 = \begin{pmatrix} 2 \\ 1 \end{pmatrix}$ , and  $v_2 = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$ . Then

$$L_A(v_1) = Av_1 = \begin{pmatrix} 7 & -10 \\ 5 & -8 \end{pmatrix} \begin{pmatrix} 2 \\ 1 \end{pmatrix} = \begin{pmatrix} 4 \\ 2 \end{pmatrix} = 2v_1$$

and

$$L_A(v_2) = Av_2 = \begin{pmatrix} 7 & -10 \\ 5 & -8 \end{pmatrix} \begin{pmatrix} 1 \\ 1 \end{pmatrix} = \begin{pmatrix} -3 \\ -3 \end{pmatrix} = -3v_2.$$

Since  $\beta = \{v_1, v_2\}$  is an (ordered) basis, both  $L_A$  and  $A$  are diagonalizable. Furthermore, if  $Q = \begin{pmatrix} 2 & 1 \\ 1 & 1 \end{pmatrix} = [I]_{\beta}^{\sigma}$ , then

$$Q^{-1}AQ = \begin{pmatrix} 2 & 0 \\ 0 & -3 \end{pmatrix} = [L_A]_{\beta}.$$

## Example

Let  $A = \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix}$ . Note that  $A^2 = O$ . If  $A$ —or equivalently  $L_A$ —were diagonalizable, then there would be a diagonal matrix  $D$  such that  $D = Q^{-1}AQ$ . But then

$D^2 = (Q^{-1}AQ)(Q^{-1}AQ) = Q^{-1}A^2Q = O$ . But if  $D = \begin{pmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{pmatrix}$ ,

then  $D^2 = \begin{pmatrix} \lambda_1^2 & 0 \\ 0 & \lambda_2^2 \end{pmatrix}$ . So  $D^2 = O$  implies  $D = O$ . But then

$A = QDQ^{-1}$  would be the zero matrix! So  $A$  is **not** diagonalizable.

## Example

Let  $T_\theta : \mathbf{R}^2 \rightarrow \mathbf{R}^2$  be rotation by  $\theta$  radians about the origin. Then as in Example 2 in §2.1,  $T_\theta = L_{A_\theta}$  where  $A_\theta = \begin{pmatrix} \cos(\theta) & -\sin(\theta) \\ \sin(\theta) & \cos(\theta) \end{pmatrix}$ . Now if  $0 < \theta < \pi$ , then it is clear from geometric considerations that  $T_\theta$  has **no eigenvectors!** (Pictures are key here.)

## Example

The previous example shows that  $A = \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix}$  has no eigenvectors **over  $\mathbf{F} = \mathbf{R}$** . (Here  $\theta = \frac{\pi}{2}$  and remember that  $T_{\frac{\pi}{2}}$  is a linear operator on the real-vector space  $\mathbf{R}^2$ .) But we could consider  $A \in M_{2 \times 2}(\mathbf{C})$ . Then  $\begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} i \\ 1 \end{pmatrix} = \begin{pmatrix} -1 \\ i \end{pmatrix} = i \begin{pmatrix} i \\ 1 \end{pmatrix}$  and  $v = \begin{pmatrix} i \\ 1 \end{pmatrix}$  is an eigenvector with eigenvalue  $i$ . Similarly,  $\begin{pmatrix} -i \\ 1 \end{pmatrix}$  is an eigenvector with eigenvalue  $-i$  and over the complex numbers  $A$  is diagonalizable with  $\begin{pmatrix} i & 0 \\ 0 & -i \end{pmatrix} = Q^{-1}AQ$  with  $Q = \begin{pmatrix} i & -i \\ 1 & 1 \end{pmatrix}$ . Fortunately, we are going to be working mostly over the reals. But we will need to think about the complex case from time to time.

## Example

Now let  $V = C^\infty(\mathbf{R})$  the subspace of the real-vector space  $\mathcal{F}(\mathbf{R}, \mathbf{R})$  of functions from  $\mathbf{R}$  to  $\mathbf{R}$  which have derivatives of all orders at every point. Define a linear operator  $T$  on  $C^\infty(\mathbf{R})$  by  $T(f) = f'$ . What are the eigenvectors and eigenvalues of  $T$ ?

## Solution

*If  $f$  is an eigenvector, then  $f' = \lambda f$  for some  $\lambda \in \mathbf{R}$ . But then we are supposed to know from back in the day that this happens if and only if  $f(x) = Ce^{\lambda x}$  for some constant  $C \in \mathbf{R}$  such that  $C \neq 0$ . (Remember that eigenvector cannot be equal to  $0_V$ !) Therefore every  $\lambda \in \mathbf{R}$  is an eigenvalue for  $T$  and the eigenvectors with eigenvalue  $\lambda$ , or eigenfunctions if you like, are multiples of  $f(x) = e^{\lambda x}$ . (Thus the nonzero constant functions are the eigenfunctions for  $\lambda = 0$ .)*

Time for a break and some questions.

# Eigenvalues are Rare

## Theorem

Suppose that  $A \in M_{n \times n}(\mathbf{F})$ . Then  $\lambda \in \mathbf{F}$  is an eigenvalue for  $A$  if and only if  $\det(A - \lambda I_n) = 0$ .

## Proof.

Suppose that  $\lambda$  is an eigenvalue for  $A$ . Then there is a nonzero vector  $v \in \mathbf{F}^n$  such that  $Av = \lambda v$ , or equivalently,  $(A - \lambda I_n)v = 0$ . Since  $v \neq 0_V$ , this implies that  $A - \lambda I_n$  is not invertible and hence  $\det(A - \lambda I_n) = 0$ . Conversely if  $\det(A - \lambda I_n) = 0$ , then  $A - \lambda I_n$  is not invertible and there is a nontrivial solution  $v$  to  $(A - \lambda I_n)x = 0$  and then  $v$  is an eigenvector with eigenvalue  $\lambda$ .  $\square$

# Example

## Example

Let  $A = \begin{pmatrix} 7 & -10 \\ 5 & -8 \end{pmatrix}$ . Then  $\det(A - \lambda I_2) = \det \begin{pmatrix} 7-\lambda & -10 \\ 5 & -8-\lambda \end{pmatrix} = (\lambda - 7)(\lambda + 8) + 50 = \lambda^2 + \lambda - 6 = (\lambda + 3)(\lambda - 2)$ . It follows that the **only** eigenvalues of  $A$  are just  $\lambda = -3$  and  $\lambda = 2$ . Of course, this was the example we looked at earlier.

# An Observation

## Proposition

Suppose  $A \in M_{n \times n}(\mathbf{F})$ . Then  $p(\lambda) = \det(A - \lambda I_n)$  is a polynomial in  $P_n(\mathbf{F})$  of the form

$$p(\lambda) = (-1)^n \lambda^n + a_{n-1} \lambda^{n-1} + \cdots + a_1 \lambda + a_0.$$

In particular,  $p$  has degree  $n$  and leading coefficient  $(-1)^n$ .

For the proof, we need a little lemma.

## Lemma

Let  $A(\lambda)$  be a  $n \times n$  matrix with entries  $A_{ij}(\lambda) \in P_1(\mathbf{F})$ . Then  $q(\lambda) = \det(A(\lambda)) \in P_n(\mathbf{F})$ .

## Proof.

This is clear if  $n = 1$ , so let  $n \geq 2$  and assume the result for  $(n-1) \times (n-1)$ -matrices. Then

$$\det(A(\lambda)) = \sum_{j=1}^n (-1)^{1+j} A_{1j}(\lambda) \det(\widetilde{(A(\lambda))}_{1j}).$$

The result follows since  $\det(\widetilde{(A(\lambda))}_{1j}) \in P_{n-1}(\mathbf{F})$  by assumption. □



# Proof of the Proposition

The proposition is easy to check if  $n \leq 2$ . So assume  $n \geq 3$  and that the result is true for  $(n-1) \times (n-1)$ -matrices. Then

$$\begin{aligned} p(\lambda) &= \det(A - \lambda I_n) \\ &= (A_{11} - \lambda) \det(\widetilde{A}_{11} - \lambda I_{n-1}) + \sum_{j=2}^n (-1)^{1+j} A_{1j} \det(\widetilde{(A - \lambda I_n)}_{1j}) \end{aligned}$$

which, by the induction hypotheses, is

$$\begin{aligned} &= (A_{11} - \lambda) ((-1)^{n-1} \lambda^{n-1} + \cdots + b_0) + \sum_{j=2}^n (-1)^{1+j} A_{1j} \det(\widetilde{(A - \lambda I_n)}_{1j}) \\ &= (-1)^n \lambda^n + p_0(\lambda) + \sum_{j=2}^n p_j(\lambda) \end{aligned}$$

where  $p_0 \in P_{n-1}(\mathbf{F})$ . Since each  $p_j \in P_{n-1}(\mathbf{F})$  by the lemma. The result now follows easily.

# Characteristic Polynomial

## Definition

If  $A \in M_{n \times n}(\mathbf{F})$ , then  $p(\lambda) = \det(A - \lambda I_n)$  is called the **characteristic polynomial** of  $A$ .

## Proposition

If  $A \in M_{n \times n}(\mathbf{F})$ , then  $A$  has at most  $n$  eigenvalues.

## Proposition

*Our technical proposition implies that the characteristic polynomial  $p(\lambda) = \det(A - \lambda I_n)$  is a polynomial of degree  $n$ . Hence  $p$  has at most  $n$  roots. (We call  $\lambda$  a root of  $p$  if and only if  $p(\lambda) = 0$ .) But our earlier theorem implies that  $\lambda$  is an eigenvalue if and only if  $\lambda$  is a root of the characteristic polynomial.*

# Finding Eigenvectors

## Remark

Let  $A \in M_{n \times n}(\mathbf{F})$ . If  $\lambda$  is a root of the characteristic polynomial  $p(\lambda) = \det(A - \lambda I_n)$ , then  $\lambda$  is an eigenvalue. The eigenvectors with eigenvalue  $\lambda$  are exactly the nontrivial solutions to  $(A - \lambda I_n)x = 0$ .

## Example

Try to diagonalize  $A = \begin{pmatrix} 1 & 2 \\ 3 & 2 \end{pmatrix}$ .

## Solution

*First compute the characteristic polynomial  $p(\lambda) = \det(A - \lambda I_n) = \det\begin{pmatrix} 1-\lambda & 2 \\ 3 & 2-\lambda \end{pmatrix} = (\lambda - 1)(\lambda - 2) - 6 = \lambda^2 - 3\lambda - 4 = (\lambda + 1)(\lambda - 4)$ .*

## Solution (Solution Continued)

Therefore the eigenvalues are just  $\lambda = -1$  and  $\lambda = 4$ . To find eigenvalues, we consider  $(A - \lambda I_2)x = 0$ . Since  $A + I_2 = \begin{pmatrix} 2 & 2 \\ 3 & 3 \end{pmatrix}$  is row equivalent to  $\begin{pmatrix} 1 & 1 \\ 0 & 0 \end{pmatrix}$ , the solutions to  $(A + I_2)x = 0$  are the same as the solutions to  $\begin{pmatrix} 1 & 1 \\ 0 & 0 \end{pmatrix}x = 0$ . (Why!?! Consider the augmented matrices.) Thus all the eigenvectors are (nonzero) multiples of  $\begin{pmatrix} -1 \\ 1 \end{pmatrix}$ . On the other hand  $A - 4I_2 = \begin{pmatrix} -3 & 2 \\ 3 & -2 \end{pmatrix}$  is row equivalent to  $\begin{pmatrix} -3 & 2 \\ 0 & 0 \end{pmatrix}$ , so the solutions to  $(A - 4I_2)x = 0$  are the same as for  $\begin{pmatrix} -3 & 2 \\ 0 & 0 \end{pmatrix}x = 0$ . Hence the eigenvectors with eigenvalue 4 are all (nonzero) multiples of  $\begin{pmatrix} 2 \\ 3 \end{pmatrix}$ . Hence  $\left\{ \begin{pmatrix} -1 \\ 1 \end{pmatrix}, \begin{pmatrix} 2 \\ 3 \end{pmatrix} \right\}$  is a basis of eigenvectors for  $A$  and if  $Q = \begin{pmatrix} -1 & 2 \\ 1 & 3 \end{pmatrix}$  then

$$Q^{-1}AQ = \begin{pmatrix} -1 & 0 \\ 0 & 4 \end{pmatrix}.$$

Note that there are many possible choices for  $Q$  here!

# A Bigger Example

## Example

Now let  $A = \begin{pmatrix} 1 & 3 & 3 \\ -3 & -5 & -3 \\ 3 & 3 & 1 \end{pmatrix}$ . Then

$p(\lambda) = \det(A - \lambda I_3) = \det \begin{pmatrix} 1 - \lambda & 3 & 3 \\ -3 & -5 - \lambda & -3 \\ 3 & 3 & 1 - \lambda \end{pmatrix}$  is given by

$$p(\lambda) = (1 - \lambda) \det \begin{pmatrix} -5 - \lambda & -3 \\ 3 & 1 - \lambda \end{pmatrix} - 3 \det \begin{pmatrix} -3 & -3 \\ 3 & 1 - \lambda \end{pmatrix} + 3 \det \begin{pmatrix} -3 & -5 - \lambda \\ 3 & 3 \end{pmatrix}$$

$$= (1 - \lambda) [(\lambda + 5)(\lambda - 1) + 9] - 3(3(\lambda - 1) + 9) \\ + 3(-9 + 3(\lambda + 5))$$

$$= (1 - \lambda)(\lambda^2 + 4\lambda + 4) - 9(\lambda + 2) + 9(\lambda + 2)$$

$$= -(\lambda - 1)(\lambda + 2)^2.$$

# Example

## Example (Continued)

Thus the eigenvalues are  $\lambda = 1$  and  $\lambda = -2$ . To find the eigenvectors for  $\lambda = -2$  we consider

$$A + 2I_3 = \begin{pmatrix} 3 & 3 & 3 \\ -3 & -3 & -3 \\ 3 & 3 & 3 \end{pmatrix} \underset{\text{row equivalent}}{\sim} \begin{pmatrix} 1 & 1 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix}.$$

Hence the eigenvectors for  $\lambda = -2$  are spanned by  $v_1 = (-1, 0, 1)$  and  $v_2 = (1, -1, 0)$ .

Similarly,

$$A - I_3 = \begin{pmatrix} 0 & 3 & 3 \\ -3 & -6 & -3 \\ 3 & 3 & 0 \end{pmatrix} \sim \begin{pmatrix} 1 & 0 & -1 \\ 0 & 1 & 1 \\ 0 & 0 & 0 \end{pmatrix}.$$

Thus the  $\lambda = 1$  eigenvectors are nonzero multiples of  $v_3 = (1, -1, 1)$ .

## Example (Continued)

Now we need to see that  $\beta = \{v_1, v_2, v_3\}$  is a basis. But if  $Q$  is the matrix with columns  $v_1$ ,  $v_2$ , and  $v_3$ , then

$$\begin{aligned}\det(Q) &= \det \begin{pmatrix} -1 & -1 & 1 \\ 0 & 1 & -1 \\ 1 & 0 & 1 \end{pmatrix} = \det \begin{pmatrix} -1 & -1 & 0 \\ 0 & 1 & 0 \\ 1 & 0 & 1 \end{pmatrix} \\ &= \det \begin{pmatrix} -1 & -1 \\ 0 & 1 \end{pmatrix} = -1 \neq 0.\end{aligned}$$

Hence  $\beta$  is a basis of eigenvectors and  $Q^{-1}AQ = \begin{pmatrix} -2 & 0 & 0 \\ 0 & -2 & 0 \\ 0 & 0 & 1 \end{pmatrix}$ .

# Enough

- 1 That is enough for today.