

# Section 5.4

## Diagonalization

## Important note about this section

In this section we discuss what it means for an  $n \times n$  matrix  $A$  to be *diagonalizable*. This term is sometimes called “diagonalizable over  $\mathbf{R}$ .”

We emphasize that any time we mention the term “diagonalizable” for a matrix  $A$  in Math 1553, all matrices involved are assumed to have only **real numbers** and the eigenvalues of  $A$  will all be real numbers.

(Side note: there is also a concept called “diagonalizable over  $\mathbf{C}$ ” which generalizes the concept of diagonalizability to cases where the matrix in question may have some eigenvalues that are not real numbers. We will discuss complex eigenvalues in section 5.5, but we do not cover diagonalizability over  $\mathbf{C}$  in Math 1553.)

# Motivation

## Difference equations

Many real-word linear algebra problems have the form:

$$v_1 = Av_0, \quad v_2 = Av_1 = A^2v_0, \quad v_3 = Av_2 = A^3v_0, \quad \dots \quad v_n = Av_{n-1} = A^n v_0.$$

This is called a **difference equation**.

Our toy example about rabbit populations had this form.

The question is, what happens to  $v_n$  as  $n \rightarrow \infty$ ?

- ▶ Taking powers of diagonal matrices is easy!
- ▶ Taking powers of *diagonalizable* matrices is still easy!
- ▶ Diagonalizing a matrix is an eigenvalue problem.

## Powers of Diagonal Matrices

If  $D$  is diagonal, then  $D^n$  is also diagonal; its diagonal entries are the  $n$ th powers of the diagonal entries of  $D$ :

$$D = \begin{pmatrix} 2 & 0 \\ 0 & -1 \end{pmatrix}, \quad D^2 = \begin{pmatrix} 4 & 0 \\ 0 & 1 \end{pmatrix}, \quad D^3 = \begin{pmatrix} 8 & 0 \\ 0 & -1 \end{pmatrix}, \quad \dots \quad D^n = \begin{pmatrix} 2^n & 0 \\ 0 & (-1)^n \end{pmatrix}.$$

$$D = \begin{pmatrix} -1 & 0 & 0 \\ 0 & \frac{1}{2} & 0 \\ 0 & 0 & \frac{1}{3} \end{pmatrix}, \quad D^2 = \begin{pmatrix} 1 & 0 & 0 \\ 0 & \frac{1}{4} & 0 \\ 0 & 0 & \frac{1}{9} \end{pmatrix}, \quad D^3 = \begin{pmatrix} -1 & 0 & 0 \\ 0 & \frac{1}{8} & 0 \\ 0 & 0 & \frac{1}{27} \end{pmatrix},$$
$$\dots \quad D^n = \begin{pmatrix} (-1)^n & 0 & 0 \\ 0 & \frac{1}{2^n} & 0 \\ 0 & 0 & \frac{1}{3^n} \end{pmatrix}$$

## Powers of Matrices that are Similar to Diagonal Ones

What if  $A$  is not diagonal?

Example

Let  $A = \begin{pmatrix} 1/2 & 3/2 \\ 3/2 & 1/2 \end{pmatrix}$ . Compute  $A^n$ , using

$$A = CDC^{-1} \quad \text{for} \quad C = \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix} \quad \text{and} \quad D = \begin{pmatrix} 2 & 0 \\ 0 & -1 \end{pmatrix}.$$

We compute:

$$A^2 = (CDC^{-1})(CDC^{-1}) = CD(C^{-1}C)DC^{-1} = CDIDC^{-1} = CD^2C^{-1}$$

$$A^3 = (CDC^{-1})(CD^2C^{-1}) = CD(C^{-1}C)D^2C^{-1} = CDID^2C^{-1} = CD^3C^{-1}$$

$\vdots$

$$A^n = CD^nC^{-1}$$

Closed formula in terms of  $n$ :  
easy to compute



Therefore

$$A^n = \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix} \begin{pmatrix} 2^n & 0 \\ 0 & (-1)^n \end{pmatrix} \frac{1}{-2} \begin{pmatrix} -1 & -1 \\ -1 & 1 \end{pmatrix} = \frac{1}{2} \begin{pmatrix} 2^n + (-1)^n & 2^n + (-1)^{n+1} \\ 2^n + (-1)^{n+1} & 2^n + (-1)^n \end{pmatrix}.$$

## Similar Matrices

### Definition

Two  $n \times n$  matrices  $A$  and  $B$  (whose entries are real numbers) are **similar** if there exists an invertible  $n \times n$  matrix  $C$  (whose entries are real numbers) such that  $A = CBC^{-1}$ .

**Fact:** if two matrices are similar then so are their powers:

$$A = CBC^{-1} \implies A^n = CB^nC^{-1}.$$

**Fact:** if  $A$  is similar to  $B$  and  $B$  is similar to  $D$ , then  $A$  is similar to  $D$ .

$$A = CBC^{-1}, \quad B = EDE^{-1} \implies A = C(EDE^{-1})C^{-1} = (CE)D(CE)^{-1}.$$

## Diagonalizable Matrices

### Definition

An  $n \times n$  matrix  $A$  is **diagonalizable** if it is similar to a diagonal matrix:

$$A = CDC^{-1} \quad \text{for } D \text{ diagonal.}$$

Important

If  $A = CDC^{-1}$  for  $D = \begin{pmatrix} d_{11} & 0 & \cdots & 0 \\ 0 & d_{22} & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & d_{nn} \end{pmatrix}$  then

$$A^k = CD^k C^{-1} = C \begin{pmatrix} d_{11}^k & 0 & \cdots & 0 \\ 0 & d_{22}^k & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & d_{nn}^k \end{pmatrix} C^{-1}.$$

So diagonalizable matrices are easy to raise to any power.

# Diagonalization

## The Diagonalization Theorem

An  $n \times n$  matrix  $A$  is diagonalizable if and only if  $A$  has  $n$  linearly independent eigenvectors in  $\mathbf{R}^n$ .

In this case,  $A = CDC^{-1}$  for

$$C = \left( \begin{array}{c|c|c|c} | & | & \cdots & | \\ \hline v_1 & v_2 & \cdots & v_n \\ \hline | & | & & | \end{array} \right) \quad D = \begin{pmatrix} \lambda_1 & 0 & \cdots & 0 \\ 0 & \lambda_2 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \lambda_n \end{pmatrix},$$

where  $v_1, v_2, \dots, v_n$  are linearly independent eigenvectors, and  $\lambda_1, \lambda_2, \dots, \lambda_n$  are the corresponding eigenvalues (in the same order).

**Corollary**  a theorem that follows easily from another theorem

An  $n \times n$  matrix with  $n$  distinct real eigenvalues is diagonalizable.

The Corollary is true because eigenvectors with distinct eigenvalues are always linearly independent. We will see later that a diagonalizable matrix need not have  $n$  distinct eigenvalues though.



## Diagonalization

### The Diagonalization Theorem

An  $n \times n$  matrix  $A$  is diagonalizable if and only if  $A$  has  $n$  linearly independent eigenvectors in  $\mathbf{R}^n$ .

In this case,  $A = CDC^{-1}$  for

$$C = \left( \begin{array}{c|c|c|c} | & | & \cdots & | \\ v_1 & v_2 & & v_n \\ | & | & & | \end{array} \right) \quad D = \begin{pmatrix} \lambda_1 & 0 & \cdots & 0 \\ 0 & \lambda_2 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \lambda_n \end{pmatrix},$$

where  $v_1, v_2, \dots, v_n$  are linearly independent eigenvectors, and  $\lambda_1, \lambda_2, \dots, \lambda_n$  are the corresponding eigenvalues (in the same order).

Note that the decomposition is not unique: you can reorder the eigenvalues and eigenvectors.

$$A = \left( \begin{array}{c|c} | & | \\ v_1 & v_2 \\ | & | \end{array} \right) \begin{pmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{pmatrix} \left( \begin{array}{c|c} | & | \\ v_1 & v_2 \\ | & | \end{array} \right)^{-1} = \left( \begin{array}{c|c} | & | \\ v_2 & v_1 \\ | & | \end{array} \right) \begin{pmatrix} \lambda_2 & 0 \\ 0 & \lambda_1 \end{pmatrix} \left( \begin{array}{c|c} | & | \\ v_2 & v_1 \\ | & | \end{array} \right)^{-1}$$

# Diagonalization

## Easy example

**Question:** What does the Diagonalization Theorem say about the matrix

$$A = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 3 \end{pmatrix}?$$

This is a triangular matrix, so the eigenvalues are the diagonal entries 1, 2, 3.

A diagonal matrix just scales the coordinates by the diagonal entries, so we can take our eigenvectors to be the unit coordinate vectors  $e_1, e_2, e_3$ . Hence the Diagonalization Theorem says

$$\begin{pmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 3 \end{pmatrix} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 3 \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}.$$

It doesn't give us anything new because the matrix was already diagonal!

A diagonal matrix  $D$  is diagonalizable! It is similar to itself:

$$D = I_n D I_n^{-1}.$$

# Diagonalization

## Example

**Problem:** Diagonalize  $A = \begin{pmatrix} 1/2 & 3/2 \\ 3/2 & 1/2 \end{pmatrix}$ .

The characteristic polynomial is

$$f(\lambda) = \lambda^2 - \text{Tr}(A)\lambda + \det(A) = \lambda^2 - \lambda - 2 = (\lambda + 1)(\lambda - 2).$$

Therefore the eigenvalues are  $-1$  and  $2$ . Let's compute some eigenvectors:

$$(A + I)x = 0 \iff \begin{pmatrix} 3/2 & 3/2 \\ 3/2 & 3/2 \end{pmatrix} x = 0 \xrightarrow{\text{rref}} \begin{pmatrix} 1 & 1 \\ 0 & 0 \end{pmatrix} x = 0$$

The parametric form is  $x = -y$ , so  $v_1 = \begin{pmatrix} -1 \\ 1 \end{pmatrix}$  is an eigenvector with eigenvalue  $-1$ .

$$(A - 2I)x = 0 \iff \begin{pmatrix} -3/2 & 3/2 \\ 3/2 & -3/2 \end{pmatrix} x = 0 \xrightarrow{\text{rref}} \begin{pmatrix} 1 & -1 \\ 0 & 0 \end{pmatrix} x = 0$$

The parametric form is  $x = y$ , so  $v_2 = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$  is an eigenvector with eigenvalue  $2$ .

The eigenvectors  $v_1, v_2$  are linearly independent, so the Diagonalization Theorem says

$$A = CDC^{-1} \quad \text{for} \quad C = \begin{pmatrix} -1 & 1 \\ 1 & 1 \end{pmatrix} \quad D = \begin{pmatrix} -1 & 0 \\ 0 & 2 \end{pmatrix}.$$

# Diagonalization

## Another example

**Problem:** Diagonalize  $A = \begin{pmatrix} 4 & -3 & 0 \\ 2 & -1 & 0 \\ 1 & -1 & 1 \end{pmatrix}$ .

The characteristic polynomial is

$$f(\lambda) = \det(A - \lambda I) = -\lambda^3 + 4\lambda^2 - 5\lambda + 2 = -(\lambda - 1)^2(\lambda - 2).$$

Therefore the eigenvalues are 1 and 2, with respective multiplicities 2 and 1.

Let's compute the 1-eigenspace:

$$(A - I)x = 0 \iff \begin{pmatrix} 3 & -3 & 0 \\ 2 & -2 & 0 \\ 1 & -1 & 0 \end{pmatrix} x = 0 \xrightarrow{\text{rref}} \begin{pmatrix} 1 & -1 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix} x = 0$$

The parametric vector form is

$$\begin{array}{l} x = y \\ y = y \\ z = z \end{array} \implies \begin{pmatrix} x \\ y \\ z \end{pmatrix} = y \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix} + z \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$$

Hence a basis for the 1-eigenspace is

$$\mathcal{B}_1 = \{v_1, v_2\} \quad \text{where} \quad v_1 = \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}, \quad v_2 = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}.$$

## Diagonalization

Another example, continued

**Problem:** Diagonalize  $A = \begin{pmatrix} 4 & -3 & 0 \\ 2 & -1 & 0 \\ 1 & -1 & 1 \end{pmatrix}$ .

Now let's compute the 2-eigenspace:

$$(A - 2I)x = 0 \iff \begin{pmatrix} 2 & -3 & 0 \\ 2 & -3 & 0 \\ 1 & -1 & -1 \end{pmatrix} x = 0 \xrightarrow{\text{rref}} \begin{pmatrix} 1 & 0 & -3 \\ 0 & 1 & -2 \\ 0 & 0 & 0 \end{pmatrix} x = 0$$

The parametric form is  $x = 3z, y = 2z$ , so an eigenvector with eigenvalue 2 is

$$v_3 = \begin{pmatrix} 3 \\ 2 \\ 1 \end{pmatrix}.$$

The eigenvectors  $v_1, v_2, v_3$  are linearly independent:  $v_1, v_2$  form a basis for the 1-eigenspace, and  $v_3$  is not contained in the 1-eigenspace. Therefore the Diagonalization Theorem says

$$A = CDC^{-1} \quad \text{for} \quad C = \begin{pmatrix} 1 & 0 & 3 \\ 1 & 0 & 2 \\ 0 & 1 & 1 \end{pmatrix} \quad D = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 2 \end{pmatrix}.$$

**Note:** In this case, there are three linearly independent eigenvectors, but only two distinct eigenvalues.

# Diagonalization

A non-diagonalizable matrix

**Problem:** Show that  $A = \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix}$  is not diagonalizable.

This is an upper-triangular matrix, so the only eigenvalue is 1. Let's compute the 1-eigenspace:

$$(A - I)x = 0 \iff \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} x = 0.$$

This is row reduced, but has only one free variable  $x$ ; a basis for the 1-eigenspace is  $\left\{ \begin{pmatrix} 1 \\ 0 \end{pmatrix} \right\}$ . So *all eigenvectors* of  $A$  are multiples of  $\begin{pmatrix} 1 \\ 0 \end{pmatrix}$ .

**Conclusion:**  $A$  has only one linearly independent eigenvector, so by the “only if” part of the diagonalization theorem,  $A$  is not diagonalizable.

## Poll

Which of the following matrices are diagonalizable, and why?

A.  $\begin{pmatrix} 1 & 2 \\ 0 & 1 \end{pmatrix}$    B.  $\begin{pmatrix} 1 & 2 \\ 0 & 2 \end{pmatrix}$    C.  $\begin{pmatrix} 2 & 1 \\ 0 & 2 \end{pmatrix}$    D.  $\begin{pmatrix} 2 & 0 \\ 0 & 2 \end{pmatrix}$

Matrix **A** is not diagonalizable: its only eigenvalue is 1, and its 1-eigenspace is spanned by  $\begin{pmatrix} 1 \\ 0 \end{pmatrix}$ .

Similarly, matrix **C** is not diagonalizable.

Matrix **B** is diagonalizable because it is a  $2 \times 2$  matrix with distinct eigenvalues.

Matrix **D** is already diagonal!

# Diagonalization

## Procedure

### How to diagonalize a matrix $A$ :

1. Find the eigenvalues of  $A$  using the characteristic polynomial.
2. For each eigenvalue  $\lambda$  of  $A$ , compute a basis  $\mathcal{B}_\lambda$  for the  $\lambda$ -eigenspace.
3. If there are fewer than  $n$  total vectors in the union of all of the eigenspace bases  $\mathcal{B}_\lambda$ , then the matrix is not diagonalizable.
4. Otherwise, the  $n$  vectors  $v_1, v_2, \dots, v_n$  in your eigenspace bases are linearly independent, and  $A = CDC^{-1}$  for

$$C = \left( \begin{array}{c|c|c|c} | & | & \cdots & | \\ v_1 & v_2 & \cdots & v_n \\ | & | & \cdots & | \end{array} \right) \quad \text{and} \quad D = \begin{pmatrix} \lambda_1 & 0 & \cdots & 0 \\ 0 & \lambda_2 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \lambda_n \end{pmatrix},$$

where  $\lambda_i$  is the eigenvalue for  $v_i$ .



# Diagonalization

## Proof

Why is the Diagonalization Theorem true?

**A diagonalizable implies A has  $n$  linearly independent eigenvectors:** Suppose  $A = CDC^{-1}$ , where  $D$  is diagonal with diagonal entries  $\lambda_1, \lambda_2, \dots, \lambda_n$ . Let  $v_1, v_2, \dots, v_n$  be the columns of  $C$ . They are linearly independent because  $C$  is invertible. So  $Ce_i = v_i$ , hence  $C^{-1}v_i = e_i$ .

$$Av_i = CDC^{-1}v_i = CDe_i = C(\lambda_i e_i) = \lambda_i Ce_i = \lambda_i v_i.$$

Hence  $v_i$  is an eigenvector of  $A$  with eigenvalue  $\lambda_i$ . So the columns of  $C$  form  $n$  linearly independent eigenvectors of  $A$ , and the diagonal entries of  $D$  are the eigenvalues.

**A has  $n$  linearly independent eigenvectors implies A is diagonalizable:** Suppose  $A$  has  $n$  linearly independent eigenvectors  $v_1, v_2, \dots, v_n$ , with eigenvalues  $\lambda_1, \lambda_2, \dots, \lambda_n$ . Let  $C$  be the invertible matrix with columns  $v_1, v_2, \dots, v_n$ . Let  $D = C^{-1}AC$ .

$$De_i = C^{-1}ACe_i = C^{-1}Av_i = C^{-1}(\lambda_i v_i) = \lambda_i C^{-1}v_i = \lambda_i e_i.$$

Hence  $D$  is diagonal, with diagonal entries  $\lambda_1, \lambda_2, \dots, \lambda_n$ . Solving  $D = C^{-1}AC$  for  $A$  gives  $A = CDC^{-1}$ .

# Algebraic Multiplicity

## Definition

The **(algebraic) multiplicity** of an eigenvalue  $\lambda$  is its multiplicity as a root of the characteristic polynomial.

This is not a very interesting notion *yet*. It will become interesting when we also define *geometric* multiplicity later.

## Example

In the rabbit population matrix,  $f(\lambda) = -(\lambda - 2)(\lambda + 1)^2$ , so the algebraic multiplicity of the eigenvalue 2 is 1, and the algebraic multiplicity of the eigenvalue  $-1$  is 2.

## Example

In the matrix  $\begin{pmatrix} 5 & 2 \\ 2 & 1 \end{pmatrix}$ ,  $f(\lambda) = (\lambda - (3 - 2\sqrt{2}))(\lambda - (3 + 2\sqrt{2}))$ , so the algebraic multiplicity of  $3 + 2\sqrt{2}$  is 1, and the algebraic multiplicity of  $3 - 2\sqrt{2}$  is 1.

## Non-Distinct Eigenvalues

### Definition

Let  $\lambda$  be an eigenvalue of a square matrix  $A$ . The **geometric multiplicity** of  $\lambda$  is the dimension of the  $\lambda$ -eigenspace.

### Theorem

Let  $\lambda$  be an eigenvalue of a square matrix  $A$ . Then

$$1 \leq (\text{the geometric multiplicity of } \lambda) \leq (\text{the algebraic multiplicity of } \lambda).$$

The proof is beyond the scope of this course.

### Corollary

Let  $\lambda$  be an eigenvalue of a square matrix  $A$ . If the algebraic multiplicity of  $\lambda$  is 1, then the geometric multiplicity is also 1: the eigenspace is a *line*.

### The Diagonalization Theorem (Alternate Form)

Let  $A$  be an  $n \times n$  matrix. The following are equivalent:

1.  $A$  is diagonalizable.
2. The sum of the geometric multiplicities of the eigenvalues of  $A$  equals  $n$ .
3. The sum of the algebraic multiplicities of the eigenvalues of  $A$  equals  $n$ , and for each eigenvalue, *the geometric multiplicity equals the algebraic multiplicity*.

# Non-Distinct Eigenvalues

## Examples

### Example

If  $A$  has  $n$  distinct eigenvalues, then the algebraic multiplicity of each equals 1, hence so does the geometric multiplicity, and therefore  $A$  is diagonalizable.

For example,  $A = \begin{pmatrix} 1/2 & 3/2 \\ 3/2 & 1/2 \end{pmatrix}$  has eigenvalues  $-1$  and  $2$ , so it is diagonalizable.

### Example

The matrix  $A = \begin{pmatrix} 4 & -3 & 0 \\ 2 & -1 & 0 \\ 1 & -1 & 1 \end{pmatrix}$  has characteristic polynomial

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

The algebraic multiplicities of 1 and 2 are 2 and 1, respectively. They sum to 3. We showed before that the geometric multiplicity of 1 is 2 (the 1-eigenspace has dimension 2). The eigenvalue 2 automatically has geometric multiplicity 1. Hence the geometric multiplicities add up to 3, so  $A$  is diagonalizable.

## Non-Distinct Eigenvalues

Another example

### Example

The matrix  $A = \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix}$  has characteristic polynomial  $f(\lambda) = (\lambda - 1)^2$ .

It has one eigenvalue 1 of algebraic multiplicity 2.

We showed before that the geometric multiplicity of 1 is 1 (the 1-eigenspace has dimension 1).

Since the geometric multiplicity is smaller than the algebraic multiplicity, the matrix is *not* diagonalizable.

## Summary

- ▶ A matrix  $A$  is **diagonalizable** if it is similar to a diagonal matrix  $D$ :  
 $A = CDC^{-1}$ .
- ▶ It is easy to take powers of diagonalizable matrices:  $A^r = CD^rC^{-1}$ .
- ▶ An  $n \times n$  matrix is diagonalizable if and only if it has  $n$  linearly independent eigenvectors  $v_1, v_2, \dots, v_n$ , in which case  $A = CDC^{-1}$  for

$$C = \left( \begin{array}{c|c|c|c} | & | & \cdots & | \\ \hline v_1 & v_2 & \cdots & v_n \\ \hline | & | & \cdots & | \end{array} \right) \quad D = \begin{pmatrix} \lambda_1 & 0 & \cdots & 0 \\ 0 & \lambda_2 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \lambda_n \end{pmatrix}.$$

- ▶ If  $A$  has  $n$  distinct eigenvalues, then it is diagonalizable.
- ▶ The **geometric multiplicity** of an eigenvalue  $\lambda$  is the dimension of the  $\lambda$ -eigenspace.
- ▶  $1 \leq (\text{geometric multiplicity}) \leq (\text{algebraic multiplicity})$ .
- ▶ An  $n \times n$  matrix is diagonalizable if and only if the sum of the geometric multiplicities is  $n$ .