## Section 6.4

 ${\sf Diagonalization}$ 

Many real-word linear algebra problems have the form:

$$v_1 = Av_0, \quad v_2 = Av_1 = A^2v_0, \quad v_3 = Av_2 = A^3v_0, \quad \dots \quad v_n = Av_{n-1} = A^nv_0.$$

This is called a difference equation.

Our toy example about rabbit populations had this form.

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

- ▶ 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 nth powers of the diagonal entries of D:

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

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

## 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

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

We compute:

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

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

$$\vdots$$

$$A^{n} = CD^{n}C^{-1}$$
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 are **similar** if there exists an invertible  $n \times n$  matrix C 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.

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

## Diagonalizable Matrices

#### Definition

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

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

$$C = \begin{pmatrix} | & | & & | \\ v_1 & v_2 & \cdots & v_n \\ | & | & & | \end{pmatrix} \qquad D = \begin{pmatrix} \lambda_1 & 0 & \cdots & 0 \\ 0 & \lambda_2 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \lambda_n \end{pmatrix},$$

where  $v_1, v_2, \ldots, v_n$  are linearly independent eigenvectors, and  $\lambda_1, \lambda_2, \ldots, \lambda_n$  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 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 this case,  $A = CDC^{-1}$  for

$$C = \begin{pmatrix} | & | & & | \\ v_1 & v_2 & \cdots & v_n \\ | & | & & | \end{pmatrix} \qquad D = \begin{pmatrix} \lambda_1 & 0 & \cdots & 0 \\ 0 & \lambda_2 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \lambda_n \end{pmatrix},$$

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

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

$$A = \begin{pmatrix} \begin{vmatrix} & & | \\ v_1 & v_2 \\ | & & | \end{pmatrix} \begin{pmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{pmatrix} \begin{pmatrix} \begin{vmatrix} & & | \\ v_1 & v_2 \\ | & & | \end{pmatrix}^{-1} = \begin{pmatrix} \begin{vmatrix} & & | \\ v_2 & v_1 \\ | & & | \end{pmatrix} \begin{pmatrix} \lambda_2 & 0 \\ 0 & \lambda_1 \end{pmatrix} \begin{pmatrix} \begin{vmatrix} & & | \\ v_2 & v_1 \\ | & & | \end{pmatrix}^{-1}$$

## 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:

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

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:

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

The parametric form is x = -y, so  $v_1 = \binom{-1}{1}$  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  $v_2 = \binom{1}{1}$  is an eigenvector with eigenvalue 2.

The eigenvectors  $\emph{v}_1, \emph{v}_2$  are linearly independent, so the Diagonalization Theorem says

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

## 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

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

Hence a basis for the 1-eigenspace is

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

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

Now let's compute the 2-eigenspace:

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

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

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

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  $\binom{1}{0}$ . So all eigenvectors of A are multiples of  $\binom{1}{0}$ .

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  $\binom{1}{0}$ .

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!

#### 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  $v_1, v_2, \ldots, v_n$  in your eigenspace bases are linearly independent, and  $A = CDC^{-1}$  for

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

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

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  $\lambda_1, \lambda_2, \ldots, \lambda_n$ . Let  $v_1, v_2, \ldots, v_n$  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  $v_1, v_2, \ldots, v_n$ , with eigenvalues  $\lambda_1, \lambda_2, \ldots, \lambda_n$ . Let C be the invertible matrix with columns  $v_1, v_2, \ldots, v_n$ . 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  $A = CDC^{-1}$ .

#### 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

 $1 \le$  (the geometric multiplicity of  $\lambda$ )  $\le$  (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 *the geometric multiplicity equals the algebraic multiplicity* of each eigenvalue.

# 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  $v_1, v_2, \ldots, v_n$ , in which case  $A = CDC^{-1}$  for

$$C = \begin{pmatrix} | & | & & | \\ v_1 & v_2 & \cdots & v_n \\ | & | & & | \end{pmatrix} \qquad D = \begin{pmatrix} \lambda_1 & 0 & \cdots & 0 \\ 0 & \lambda_2 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \lambda_n \end{pmatrix}.$$

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