## 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 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.)

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:

## 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:

$$A^2 =$$

$$A^3 =$$

$$A^n =$$

Therefore

$$A^n =$$

#### 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.

## 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.

## 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

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

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 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.

#### 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

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

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

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

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

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

# Diagonalization Example

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

Another example

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

Another example, continued

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

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

A non-diagonalizable matrix

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

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

#### 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  $v_i$ .

## Diagonalization Proof

Why is the Diagonalization Theorem true?

## 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

 $1 \le$  (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 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

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

- ▶ 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  $n \times n$  matrix is diagonalizable if and only if the sum of the geometric multiplicities is n.