

## Section 5.2

### The Characteristic Equation

# The Characteristic Polynomial

Last section we learn that for a square matrix  $A$ :

$$\begin{aligned}\lambda \text{ is an eigenvalue of } A &\iff Ax = \lambda x \text{ has a nontrivial solution} \\ &\iff (A - \lambda I)x = 0 \text{ has a nontrivial solution} \\ &\iff A - \lambda I \text{ is not invertible} \\ &\iff \det(A - \lambda I) = 0.\end{aligned}$$

## Compute Eigenvalues

The *eigenvalues* of  $A$  are **the roots** of  $\det(A - \lambda I)$

## Definition

Let  $A$  be a square matrix. The **characteristic polynomial** of  $A$  is

$$f(\lambda) = \det(A - \lambda I).$$

The **characteristic equation** of  $A$  is the equation

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

# The Characteristic Polynomial

## Example

**Question:** What are the eigenvalues of

$$A = \begin{pmatrix} 5 & 2 \\ 2 & 1 \end{pmatrix}?$$

# The Characteristic Polynomial

## Example

### Definition

The **trace** of a square matrix  $A$  is  $\text{Tr}(A) = \text{sum of the diagonal entries of } A$ .

What do you notice about: the characteristic polynomial of  $A = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$ ?

### Shortcut

The characteristic polynomial of a  $2 \times 2$  matrix  $A$  is

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

# The Characteristic Polynomial

## Example

**Question:** What are the eigenvalues of the *rabbit population matrix*

$$A = \begin{pmatrix} 0 & 6 & 8 \\ \frac{1}{2} & 0 & 0 \\ 0 & \frac{1}{2} & 0 \end{pmatrix}?$$

# Algebraic Multiplicity

## Definition

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

## Example

In the rabbit population matrix,  $f(\lambda) = -(\lambda - 2)(\lambda + 1)^2$ . The algebraic multiplicities are

$$\lambda = \begin{cases} 2 & \text{multiplicity 1} \\ -1 & \text{multiplicity 2} \end{cases}$$

## Example

In the matrix  $\begin{pmatrix} 5 & 2 \\ 2 & 1 \end{pmatrix}$ ,  $f(\lambda) = (\lambda - (3 + 2\sqrt{2}))(\lambda - (3 - 2\sqrt{2}))$ . The

algebraic multiplicities are  $\lambda = \begin{cases} 3 + 2\sqrt{2} & \text{alg. multiplicity 1} \\ 3 - 2\sqrt{2} & \text{alg. multiplicity 1} \end{cases}$

# Multiplicities

## Theorem

If  $A$  is an  $n \times n$  matrix, the **characteristic polynomial**

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

is a *polynomial* of degree  $n$ , and its *roots are the eigenvalues* of  $A$ :

$$f(\lambda) = (-1)^n \lambda^n + a_{n-1} \lambda^{n-1} + a_{n-2} \lambda^{n-2} + \cdots + a_1 \lambda + a_0.$$

### Complex numbers

If you **count the eigenvalues** of  $A$ , with their algebraic multiplicities, depending on *whether you allow complex eigenvalues*, you will get :

- ▶ **Do allow** complex numbers: *Always  $n$ .*
- ▶ **Only real** numbers: Always at most  $n$ , but *sometimes less.*

# Similarity

## Definition

Two  $n \times n$  matrices  $A$  and  $B$  are **similar** if there is an invertible  $n \times n$  matrix  $C$  such that

$$A = CBC^{-1}.$$

### The intuition

$C$  keeps record of a basis  $\mathcal{C} = \{v_1, \dots, v_n\}$  of  $\mathbf{R}^n$ .

$B$  transforms the  **$\mathcal{C}$ -coordinates** of  $x$ :  $B[x]_{\mathcal{C}} = [Ax]_{\mathcal{C}}$   
in *the same way that*  $A$  transforms the **standard coordinates** of  $x$



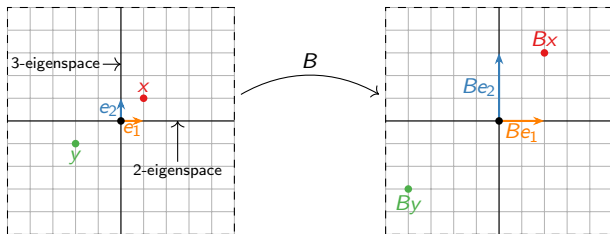
# Similarity

## Example

$$A = \begin{pmatrix} 1 & 2 \\ -1 & 4 \end{pmatrix} \quad B = \begin{pmatrix} 2 & 0 \\ 0 & 3 \end{pmatrix} \quad C = \begin{pmatrix} 2 & 1 \\ 1 & 1 \end{pmatrix} \quad \implies \quad A = CBC^{-1}.$$

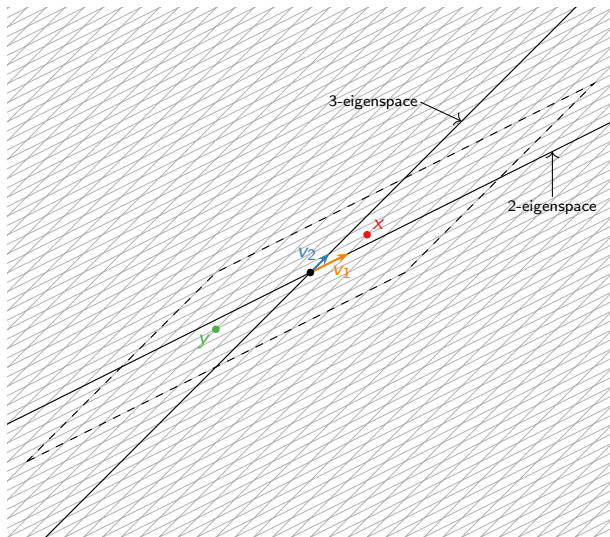
What does  $B$  do geometrically? **Scaling:**  $x$ -direction by 2 and  $y$ -direction by 3.

$B$  acting on the usual coordinates



Now  $A$  will do to the standard coordinates what  $B$  does to the  $\mathcal{C}$ -coordinates, where  $\mathcal{C} = \left\{ \begin{pmatrix} 2 \\ 1 \end{pmatrix}, \begin{pmatrix} 1 \\ 1 \end{pmatrix} \right\}$ .

# From $\mathcal{C}$ -coordinates to standard coordinates



$$\left. \begin{aligned} v_1 &= \begin{pmatrix} 2 \\ 1 \end{pmatrix} \\ v_2 &= \begin{pmatrix} 1 \\ 1 \end{pmatrix} \end{aligned} \right\} \text{vectors in } \mathcal{C}$$

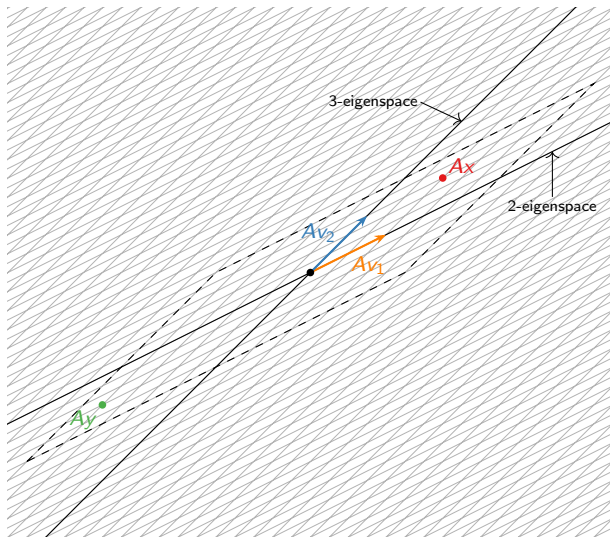
$$[x]_{\mathcal{C}} = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

$$x =$$

$$[y]_{\mathcal{C}} = \begin{pmatrix} -2 \\ -1 \end{pmatrix}$$

$$y =$$

$A$  does to the usual coordinates what  $B$  does to the  $\mathcal{C}$ -coordinates



$$Av_1 =$$

$$Av_2 =$$

$$B[x]_{\mathcal{C}} = [Ax]_{\mathcal{C}}$$

$$Ax =$$

$$B[y]_{\mathcal{C}} = [Ay]_{\mathcal{C}}$$

$$Ay =$$

Check:

# Similar Matrices Have the Same Characteristic Polynomial

## Fact

If  $A$  and  $B$  are **similar**, then they have the *same characteristic polynomial*.

**Consequence:** similar matrices have the *same eigenvalues*! Though different eigenvectors in general.

**Why?** Suppose  $A = CBC^{-1}$ . We can show that  $\det(A - \lambda I) = \det(B - \lambda I)$ .

# Similarity Caveats

## Warning

1. Matrices with the *same eigenvalues* need not be similar.  
For instance,

$$\begin{pmatrix} 2 & 1 \\ 0 & 2 \end{pmatrix} \quad \text{and} \quad \begin{pmatrix} 2 & 0 \\ 0 & 2 \end{pmatrix}$$

both only have the eigenvalue 2, but they are *not similar*.

2. Similarity is *lost in row equivalence*.  
For instance,

$$\begin{pmatrix} 2 & 1 \\ 0 & 2 \end{pmatrix} \quad \text{and} \quad \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

are row equivalent, but they have *different eigenvalues*.

# Once more: The Invertible Matrix Theorem

## Addenda

We have a couple of new ways of saying “ $A$  is invertible” now:

### The Invertible Matrix Theorem

Let  $A$  be a square  $n \times n$  matrix, and let  $T: \mathbf{R}^n \rightarrow \mathbf{R}^n$  be the linear transformation  $T(x) = Ax$ . *The following statements are equivalent.*

1.  $A$  is **invertible**.
2.  $T$  is invertible.
3.  $A$  is row equivalent to  $I_n$ .
4.  $A$  has  $n$  pivots.
5.  $Ax = 0$  has only the trivial solution.
6. The columns of  $A$  are linearly independent.
7.  $T$  is one-to-one.
8.  $Ax = b$  is consistent for all  $b$  in  $\mathbf{R}^n$ .
9. The columns of  $A$  span  $\mathbf{R}^n$ .
10.  $T$  is onto.
11.  $A$  has a left inverse (there exists  $B$  such that  $BA = I_n$ ).
12.  $A$  has a right inverse (there exists  $B$  such that  $AB = I_n$ ).
13.  $A^T$  is invertible.
14. The columns of  $A$  form a basis for  $\mathbf{R}^n$ .
15.  $\text{Col } A = \mathbf{R}^n$ .
16.  $\dim \text{Col } A = n$ .
17.  $\text{rank } A = n$ .
18.  $\text{Nul } A = \{0\}$ .
19.  $\dim \text{Nul } A = 0$ .
19. The determinant of  $A$  is *not equal to zero*.
20. The number 0 is *not an eigenvalue* of  $A$ .