Recall: an  $n \times n$  matrix A is diagonalizable if it is similar to a diagonal matrix:

$$A = CDC^{-1} \quad \text{for} \quad D = \begin{pmatrix} \lambda_1 & 0 & \cdots & 0 \\ 0 & \lambda_2 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \lambda_n \end{pmatrix}.$$

It is easy to take powers of diagonalizable matrices:

$$A^{i} = CD^{i}C^{-1} = C egin{pmatrix} \lambda_{1}^{i} & 0 & \cdots & 0 \ 0 & \lambda_{2}^{i} & \cdots & 0 \ \vdots & \vdots & \ddots & \vdots \ 0 & 0 & \cdots & \lambda_{n}^{i} \end{pmatrix} C^{-1}.$$

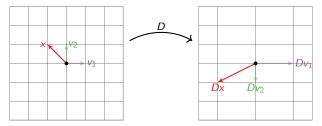
We begin today by discussing the geometry of diagonalizable matrices.

# Geometry of Diagonal Matrices

A diagonal matrix  $D = \begin{pmatrix} 2 & 0 \\ 0 & -1 \end{pmatrix}$  just scales the coordinate axes:

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

This is easy to visualize:



$$\mathbf{x} = \begin{pmatrix} -1 \\ 1 \end{pmatrix} \implies D\mathbf{x} = \begin{pmatrix} -2 \\ -1 \end{pmatrix}.$$

## Geometry of Diagonal izable Matrices

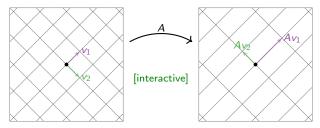
We had this example last time:  $A = CDC^{-1}$  for

$$A = \begin{pmatrix} 1/2 & 3/2 \\ 3/2 & 1/2 \end{pmatrix} \qquad D = \begin{pmatrix} 2 & 0 \\ 0 & -1 \end{pmatrix} \qquad C = \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix}$$

The eigenvectors of A are  $v_1 = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$  and  $v_2 = \begin{pmatrix} 1 \\ -1 \end{pmatrix}$  with eigenvalues 2 and -1. The eigenvectors form a *basis* for  $\mathbf{R}^2$  because they're linearly independent. Any vector can be written as a linear combination of basis vectors:

$$x = a_1 v_1 + a_2 v_2 \implies Ax =$$

Conclusion: A scales the " $v_1$ -direction" by 2 and the " $v_2$ -direction" by -1.

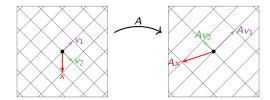


# Geometry of Diagonal *izable* Matrices

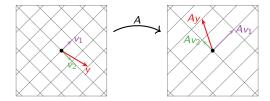
Continued

Example: 
$$x = \begin{pmatrix} 0 \\ -2 \end{pmatrix} = -1v_1 + 1v_2$$

$$A_{\mathbf{X}} = -1Av_1 + 1Av_2 = -2v_1 + -1v_2$$



Example:  $y = \frac{1}{2} {5 \choose -3} = \frac{1}{2} v_1 + 2v_2$  $Ay = \frac{1}{2} Av_1 + 2Av_2 = 1v_1 + -2v_2$ 



# Dynamics of Diagonalizable Matrices

We motivated diagonalization by taking powers:

$$A^{i} = CD^{i}C^{-1} = C \begin{pmatrix} \lambda_{1}^{i} & 0 & \cdots & 0 \\ 0 & \lambda_{2}^{i} & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \lambda_{n}^{i} \end{pmatrix} C^{-1}.$$

This lets us compute powers of matrices easily. How to visualize this?

$$A^n v = A(A(A \cdots (Av)) \cdots)$$

Multiplying a vector v by  $A^n$  means repeatedly multiplying by A.

# Dynamics of Diagonalizable Matrices Example

$$A = \frac{1}{10} \begin{pmatrix} 11 & 6 \\ 9 & 14 \end{pmatrix} = CDC^{-1} \quad \text{for} \quad C = \begin{pmatrix} 2/3 & 1 \\ -1 & 1 \end{pmatrix} \qquad D = \begin{pmatrix} 2 & 0 \\ 0 & 1/2 \end{pmatrix}.$$

Eigenvectors of A are  $v_1 = \binom{2/3}{-1}$  and  $v_2 = \binom{1}{1}$  with eigenvalues 2 and 1/2.

$$A(a_1v_1 + a_2v_2) = 2a_1v_1 + \frac{1}{2}a_2v_2$$

$$A^2(a_1v_1 + a_2v_2) = 4a_1v_1 + \frac{1}{4}a_2v_2$$

$$A^3(a_1v_1 + a_2v_2) = 8a_1v_1 + \frac{1}{8}a_2v_2$$

$$\vdots$$

$$A^n(a_1v_1 + a_2v_2) = 2^na_1v_1 + \frac{1}{2^n}a_2v_2$$

What does repeated application of A do geometrically?

It makes the " $v_1$ -coordinate" very big, and the " $v_2$ -coordinate" very small.

[interactive]

## Dynamics of Diagonalizable Matrices

Another Example

$$A = \frac{1}{6} \begin{pmatrix} 5 & -1 \\ -2 & 4 \end{pmatrix} = CDC^{-1} \quad \text{for} \quad C = \begin{pmatrix} -1 & 1/2 \\ 1 & 1 \end{pmatrix} \qquad D = \begin{pmatrix} 1 & 0 \\ 0 & 1/2 \end{pmatrix}.$$

Eigenvectors of A are  $v_1 = \binom{-1}{1}$  and  $v_2 = \binom{1/2}{1}$  with eigenvalues 1 and 1/2.

$$A(a_1v_1 + a_2v_2) = a_1v_1 + \frac{1}{2}a_2v_2$$

$$A^2(a_1v_1 + a_2v_2) = a_1v_1 + \frac{1}{4}a_2v_2$$

$$A^3(a_1v_1 + a_2v_2) = a_1v_1 + \frac{1}{8}a_2v_2$$

$$\vdots$$

$$A^n(a_1v_1 + a_2v_2) = a_1v_1 + \frac{1}{2^n}a_2v_2$$

What does repeated application of A do geometrically?

It "sucks everything into the 1-eigenspace."

[interactive]

# Dynamics of Diagonalizable Matrices $_{\mbox{Poll}}$

$$A = \frac{1}{30} \begin{pmatrix} 12 & 2 \\ 3 & 13 \end{pmatrix} = CDC^{-1} \quad \text{for} \quad C = \begin{pmatrix} 2/3 & -1 \\ 1 & 1 \end{pmatrix} \qquad D = \begin{pmatrix} 1/2 & 0 \\ 0 & 1/3 \end{pmatrix}.$$

# Section 5.5

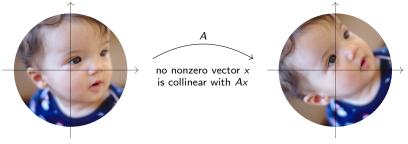
**Complex Eigenvalues** 

# A Matrix with No Eigenvectors

Consider the matrix for the linear transformation for rotation by  $\pi/4$  in the plane. The matrix is:

 $A = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & -1 \\ 1 & 1 \end{pmatrix}.$ 

This matrix has no eigenvectors, as you can see geometrically: [interactive]



or algebraically:

$$f(\lambda) = \lambda^2 - \operatorname{Tr}(A) \lambda + \det(A) = \lambda^2 - \sqrt{2} \lambda + 1 \implies \lambda = \frac{\sqrt{2} \pm \sqrt{-2}}{2}.$$

It makes us sad that -1 has no square root. If it did, then  $\sqrt{-2} = \sqrt{2} \cdot \sqrt{-1}$ .

Mathematician's solution: we're just not using enough numbers! We're going to declare by *fiat* that there exists a square root of -1.

## Definition

The number *i* is defined such that  $i^2 = -1$ .

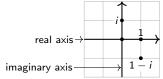
Once we have i, we have to allow numbers like a + bi for real numbers a, b.

## Definition

A complex number is a number of the form a + bi for a, b in **R**. The set of all complex numbers is denoted **C**.

Note **R** is contained in **C**: they're the numbers a + 0i.

We can identify **C** with  $\mathbf{R}^2$  by  $a + bi \leftrightarrow {a \choose b}$ . So when we draw a picture of **C**, we draw the plane:



#### Addition:

Multiplication:

Complex conjugation:  $\overline{a + bi} = a - bi$  is the complex conjugate of a + bi. Check:  $\overline{z + w} = \overline{z} + \overline{w}$  and  $\overline{zw} = \overline{z} \cdot \overline{w}$ .

Absolute value:  $|a + bi| = \sqrt{a^2 + b^2}$ . This is a *real* number. Note:  $(a + bi)(\overline{a + bi}) = (a + bi)(a - bi) = a^2 - (bi)^2 = a^2 + b^2$ . So  $|z| = \sqrt{z\overline{z}}$ . Check:  $|zw| = |z| \cdot |w|$ .

Division by a nonzero real number:  $\frac{a+bi}{c} = \frac{a}{c} + \frac{b}{c}i$ . Division by a nonzero complex number:  $\frac{z}{w} = \frac{z\overline{w}}{w\overline{w}} = \frac{z\overline{w}}{|w|^2}$ .

Example:

$$\frac{1+i}{1-i} =$$

Real and imaginary part:  $\operatorname{Re}(a + bi) = a$   $\operatorname{Im}(a + bi) = b$ .

The whole point of using complex numbers is to solve polynomial equations. It turns out that they are enough to find all solutions of all polynomial equations:

## Fundamental Theorem of Algebra

Every polynomial of degree n has exactly n complex roots, counted with multiplicity.

Equivalently, if  $f(x) = x^n + a_{n-1}x^{n-1} + \cdots + a_1x + a_0$  is a polynomial of degree *n*, then

$$f(x) = (x - \lambda_1)(x - \lambda_2) \cdots (x - \lambda_n)$$

for (not necessarily distinct) complex numbers  $\lambda_1, \lambda_2, \ldots, \lambda_n$ .

#### Important

If f is a polynomial with *real* coefficients, and if  $\lambda$  is a complex root of f, then so is  $\overline{\lambda}$ :

$$0 = \overline{f(\lambda)} = \overline{\lambda^n + a_{n-1}\lambda^{n-1} + \dots + a_1\lambda} + \overline{a_0}$$
$$= \overline{\lambda}^n + a_{n-1}\overline{\lambda}^{n-1} + \dots + a_1\overline{\lambda} + \overline{a_0} = f(\overline{\lambda})$$

Therefore complex roots of real polynomials come in conjugate pairs.

# The Fundamental Theorem of Algebra Examples

**Degree 2**: The quadratic formula gives you the (real or complex) roots of any degree-2 polynomial:

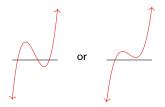
$$f(x) = x^2 + bx + c \implies x = \frac{-b \pm \sqrt{b^2 - 4c}}{2}.$$

For instance, if  $f(\lambda) = \lambda^2 - \sqrt{2}\lambda + 1$  then

 $\lambda =$ 

Note the roots are complex conjugates if b, c are real.

Degree 3: A real cubic polynomial has either three real roots, or one real root and a conjugate pair of complex roots. The graph looks like:



respectively.

# A Matrix with an Eigenvector

Every matrix is guaranteed to have complex eigenvalues and eigenvectors. Using rotation by  $\pi/4$  from before:

$$A = rac{1}{\sqrt{2}} egin{pmatrix} 1 & -1 \ 1 & 1 \end{pmatrix}$$
 has eigenvalues  $\lambda = rac{1\pm i}{\sqrt{2}}.$ 

Let's compute an eigenvector for  $\lambda = (1 + i)/\sqrt{2}$ :

A similar computation shows that an eigenvector for  $\lambda = (1 - i)/\sqrt{2}$  is  $\binom{-i}{1}$ . So is  $i\binom{-i}{1} = \binom{1}{i}$  (you can scale by *complex* numbers).

# Conjugate Eigenvectors

For 
$$A = rac{1}{\sqrt{2}} \begin{pmatrix} 1 & -1 \\ 1 & 1 \end{pmatrix}$$
,

the eigenvalue 
$$\frac{1+i}{\sqrt{2}}$$
 has eigenvector  $\begin{pmatrix} i\\1 \end{pmatrix}$ .  
the eigenvalue  $\frac{1-i}{\sqrt{2}}$  has eigenvector  $\begin{pmatrix} -i\\1 \end{pmatrix}$ .

Do you notice a pattern?

#### Fact

Let A be a real square matrix. If  $\lambda$  is a complex eigenvalue with eigenvector v, then  $\overline{\lambda}$  is an eigenvalue with eigenvector  $\overline{v}$ .

Why?

$$Av = \lambda \implies A\overline{v} = \overline{Av} = \overline{\lambda v} = \overline{\lambda v}.$$

Both eigenvalues and eigenvectors of real square matrices occur in conjugate pairs.

# $2 \times 2$ eigenvector trick

Suppose A is a 2  $\times$  2 matrix and  $\lambda$  is any eigenvalue of A. Then

$$A - \lambda I_2 = \begin{pmatrix} z & w \\ (*) & (*) \end{pmatrix} \implies \begin{pmatrix} -w \\ z \end{pmatrix}$$

is an eigenvector of A corresponding to the eigenvalue  $\lambda$ .

In the previous example, 
$$\lambda = \frac{1+i}{\sqrt{2}}$$
 was an eigenvalue of  $A = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & -1 \\ 1 & 1 \end{pmatrix}$   
and

$$A - \lambda I = \begin{pmatrix} -\frac{i}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & -\frac{i}{\sqrt{2}} \end{pmatrix} = \begin{pmatrix} z & w \\ (*) & (*) \end{pmatrix}.$$

So an eigenvector of A corresponding to  $\lambda$  is

$$v = \begin{pmatrix} -w \\ z \end{pmatrix} = \begin{pmatrix} \frac{1}{\sqrt{2}} \\ -\frac{i}{\sqrt{2}} \end{pmatrix}$$

.

This was much faster than doing the full  $2 \times 2$  row reduction in the previous example, and it agrees with our answer.

# A 3 $\times$ 3 Example

Find the eigenvalues and eigenvectors of

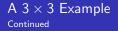
$$A = \begin{pmatrix} \frac{4}{5} & -\frac{3}{5} & 0\\ \frac{3}{5} & \frac{4}{5} & 0\\ 0 & 0 & 2 \end{pmatrix}.$$

The characteristic polynomial is

We computed the roots of this polynomial (times 5) before:

$$\lambda = -2, \quad \frac{4+3i}{5}, \quad \frac{4-3i}{5}$$

We eyeball an eigenvector with eigenvalue 2 as (0, 0, 1).



$$A = \begin{pmatrix} \frac{4}{5} & -\frac{3}{5} & 0\\ \frac{3}{5} & \frac{4}{5} & 0\\ 0 & 0 & 2 \end{pmatrix}$$

To find the other eigenvectors, we row reduce:

# Summary

- > Diagonal matrices are easy to understand geometrically.
- Diagonalizable matrices behave like diagonal matrices, except with respect to a basis of eigenvectors.
- Repeatedly multiplying by a matrix leads to fun pictures.
- One can do arithmetic with complex numbers just like real numbers: add, subtract, multiply, divide.
- ► An n × n matrix always exactly has complex n eigenvalues, counted with (algebraic) multiplicity.
- The complex eigenvalues and eigenvectors of a *real* matrix come in complex conjugate pairs:

$$Av = \lambda v \implies A\overline{v} = \overline{\lambda}\overline{v}.$$