

# Announcements Mar 28

- WebWork 5.2 and 5.3 due Thursday
- Quiz 8 on 5.2 and 5.3 on Friday
- Homework 7 due Friday April 8 *HW 8 due April 1.*
- Midterm 3 in class *Friday April 8* on *Chapter 5*
- Office Hours Tuesday and Wednesday 2-3, after class, and by appt in Skiles 244 *or 236*
- LA Office Hours: Scott Mon 12-1, Yashvi Mon 2-3, Shivang Tue 5-6, Baishen Wed 4-5, Matt Thu 3-4
- Math Lab, Clough 280
  - Regular hours: Mon/Wed 11-5 and Tue/Thu 11-5
  - Math 1553 hours: Mon-Thu 5-6 and Tue/Thu 11-12
  - LA hours: Matt Tue 11-12, Scott Tue 5-6, Baishen Thu 11-12, Yashvi/Shivang Thu 5-6

# Section 5.3

## Diagonalization

## 5.3 Diagonalization

### Outline

- Taking powers of diagonal matrices is easy
- Taking powers of diagonalizable matrices is still easy
- Algebraic multiplicity vs geometric multiplicity vs diagonalizability
- Application: networks

$$(\lambda - 1)^2$$

## Powers of diagonal matrices

We have seen that it is useful to take powers of matrices: for instance in computing rabbit populations.

$$v, Av, A^2v, A^3v$$

If  $A$  is diagonal,  $A^k$  is easy to compute. For example:

$$\begin{pmatrix} 2 & 0 \\ 0 & 3 \end{pmatrix}^{10} = \begin{pmatrix} 2^{10} & 0 \\ 0 & 3^{10} \end{pmatrix}$$

## Powers of matrices that are similar to diagonal ones

What if  $A$  is not diagonal? Suppose we need to compute

$$\begin{pmatrix} 1 & 2 \\ -1 & 4 \end{pmatrix}^{10}$$

What would we do?

$A$  similar to  $B$ :  
 $A = CBC^{-1}$

Earlier in the notes, we saw this matrix is similar to a diagonal one:

$$\boxed{\begin{pmatrix} 1 & 2 \\ -1 & 4 \end{pmatrix}} = \begin{pmatrix} 2 & 1 \\ 1 & 1 \end{pmatrix} \boxed{\begin{pmatrix} 2 & 0 \\ 0 & 3 \end{pmatrix}} \begin{pmatrix} 2 & 1 \\ 1 & 1 \end{pmatrix}^{-1}$$

$A = C B C^{-1}$

"diagonalization"

So...

$$A^2 = (C B C^{-1})(C B C^{-1}) = C B^2 C^{-1}$$

$$A^{10} = C B^{10} C^{-1} = \begin{pmatrix} 2 & 1 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} 2^{10} & 0 \\ 0 & 3^{10} \end{pmatrix} \begin{pmatrix} 1 & -1 \\ -1 & 2 \end{pmatrix}$$

$$\boxed{A^3 = C B C^{-1} C B C^{-1} C B C^{-1} = C B^3 C^{-1}}$$

# Diagonalization

Suppose  $A$  is  $n \times n$ . We say that  $A$  is **diagonalizable** if it is similar to a diagonal matrix:

$$A = CDC^{-1}$$

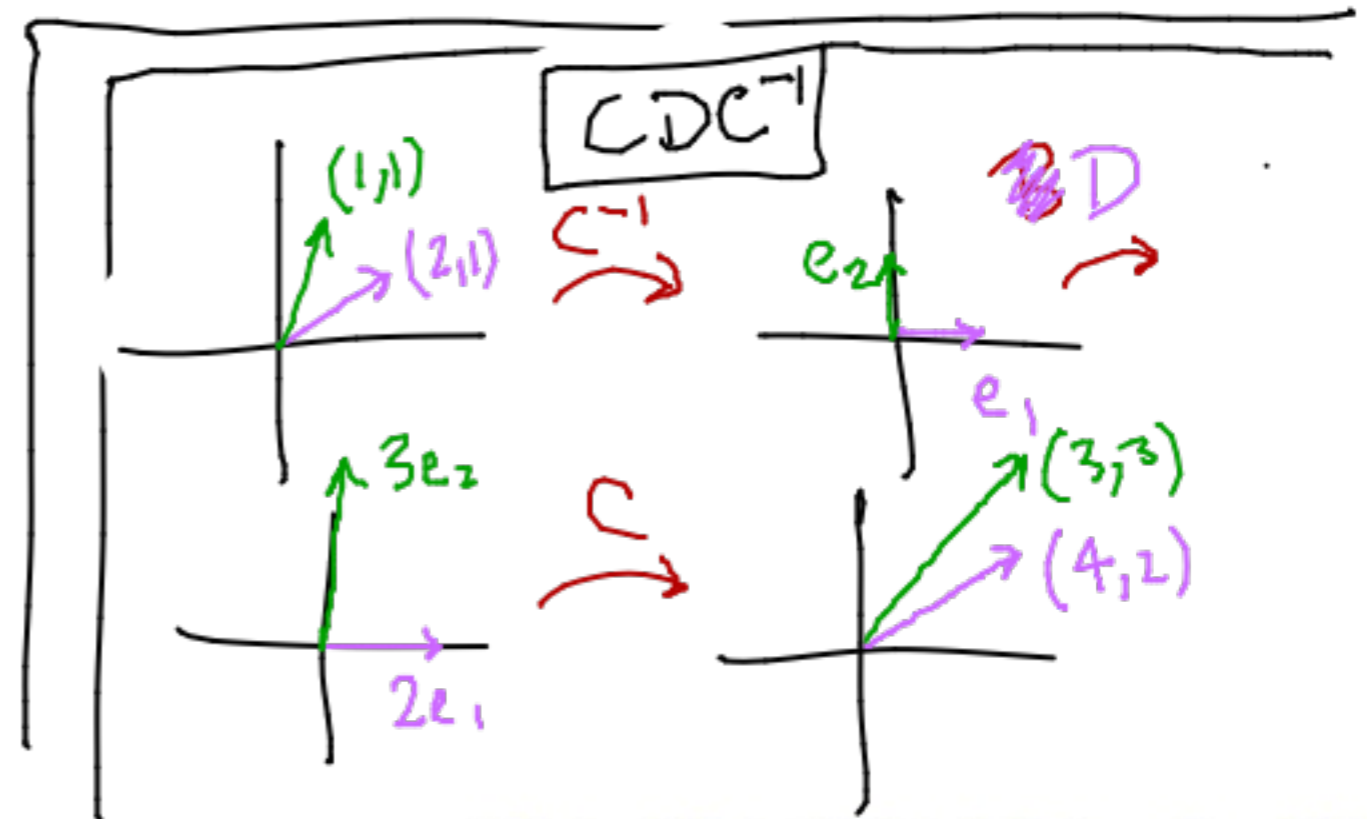
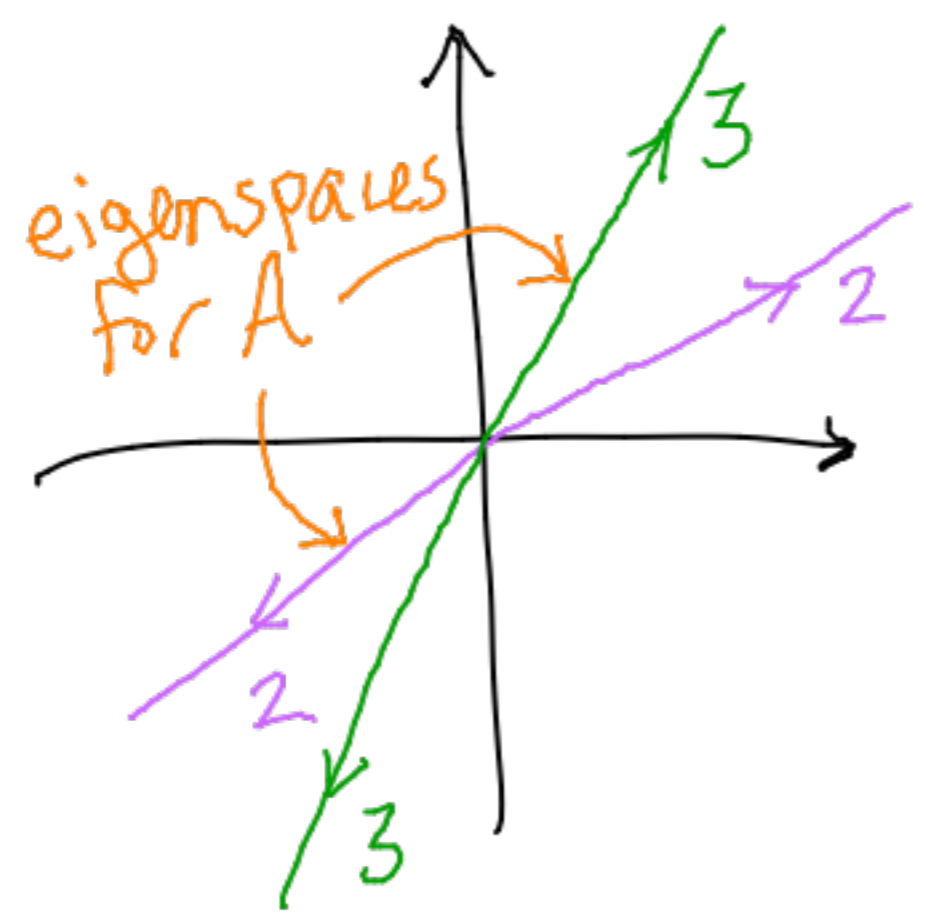
$D = \text{diagonal}$

How does this factorization of  $A$  help describe what  $A$  **does** to  $\mathbb{R}^n$ ?

$A \rightarrow \begin{pmatrix} 1 & 2 \\ -1 & 4 \end{pmatrix} = \begin{pmatrix} 2 & 1 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} 2 & 0 \\ 0 & 3 \end{pmatrix} \begin{pmatrix} 2 & 1 \\ 1 & 1 \end{pmatrix}^{-1}$

*(Handwritten annotations: The matrix  $\begin{pmatrix} 2 & 1 \\ 1 & 1 \end{pmatrix}$  is circled in orange and labeled "eigenvectors". The diagonal matrix  $\begin{pmatrix} 2 & 0 \\ 0 & 3 \end{pmatrix}$  is underlined in orange and labeled "eigenvalues".)*

$C(e_1) = \begin{pmatrix} 2 \\ 1 \end{pmatrix}$   
 $C(e_2) = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$   
 $C^{-1}\begin{pmatrix} 2 \\ 1 \end{pmatrix} = e_1$   
 $C^{-1}\begin{pmatrix} 1 \\ 1 \end{pmatrix} = e_2$



# Diagonalization

**Theorem.**  $A$  is diagonalizable  $\Leftrightarrow A$  has  $n$  linearly independent eigenvectors.

In this case

$$A = \overset{C}{(v_1 \ v_2 \ \cdots \ v_n)} \overset{D}{\begin{pmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_n \end{pmatrix}} \overset{C^{-1}}{(v_1 \ v_2 \ \cdots \ v_n)^{-1}}$$

$\begin{pmatrix} 2 & 1 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} 2 & 0 \\ 0 & 3 \end{pmatrix} \begin{pmatrix} 2 & 1 \\ 1 & 1 \end{pmatrix}^{-1}$

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

Why?

# Diagonalization

Fact. If  $A$  is diagonalizable, bases for the eigenspaces give a basis for  $\mathbb{R}^n$ .

Why?



# Example

Diagonalize if possible.

Eigenvalues

$$\begin{pmatrix} 2 & 6 \\ 0 & -1 \end{pmatrix}$$
$$\det \begin{pmatrix} 2-\lambda & 6 \\ 0 & -1-\lambda \end{pmatrix} \rightsquigarrow \lambda = 2, -1$$

distinct.

eigenvector

eigenvector

Eigenvectors

$$\boxed{\lambda = 2} \begin{pmatrix} 0 & 6 \\ 0 & -3 \end{pmatrix} \rightsquigarrow \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} \rightsquigarrow \text{eigen vector} \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$

$$\boxed{\lambda = -1} \begin{pmatrix} 3 & 6 \\ 0 & 0 \end{pmatrix} \rightsquigarrow \begin{pmatrix} 1 & 2 \\ 0 & 0 \end{pmatrix} \rightsquigarrow \text{eigen v.} \begin{pmatrix} -2 \\ 1 \end{pmatrix}$$

$$x_1 + 2x_2 = 0$$
$$x_2 \text{ free, } x_1 = -2x_2$$

Finally:

$$\begin{pmatrix} 2 & 6 \\ 0 & -1 \end{pmatrix} = \begin{pmatrix} 1 & -2 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 2 & 0 \\ 0 & -1 \end{pmatrix} \begin{pmatrix} 1 & -2 \\ 0 & 1 \end{pmatrix}^{-1}$$

## Example

Diagonalize if possible.

$$\begin{pmatrix} 3 & 1 \\ 0 & 3 \end{pmatrix}$$

Another example

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

$\lambda = 3$  (alg. mult 2) eigenvectors  $(1, 0), (0, 1)$

eigenvalues: 3 (alg mult. = 2)

eigenvectors:  $\begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} \rightsquigarrow \begin{pmatrix} 1 \\ 0 \end{pmatrix}$

only eigenvector  
up to scale

$\rightsquigarrow$  not diagonalizable.

## More Examples

Diagonalize if possible.

$$\begin{pmatrix} 0 & 1 \\ 1 & 1 \end{pmatrix}, \quad \begin{pmatrix} a & b \\ b & a \end{pmatrix}, \quad \begin{pmatrix} 1 & 0 & 2 \\ 0 & 1 & 0 \\ 2 & 0 & 1 \end{pmatrix}, \quad \begin{pmatrix} 2 & 0 & 0 \\ 1 & 2 & 1 \\ -1 & 0 & 1 \end{pmatrix}$$

✓

$$\begin{pmatrix} 2 & 0 \\ 0 & 3 \end{pmatrix}^{-1} = \begin{pmatrix} 1/2 & 0 \\ 0 & 1/3 \end{pmatrix}$$

## Poll

Which are true?

1. if  $A$  is diagonalizable then  $A^2$  is
2. if  $A$  is diagonalizable then  $A^{-1}$  is
3. if  $A^2$  is diagonalizable then  $A$  is
4. if  $A$  is diagonalizable and  $B$  is similar to  $A$  then  $B$  is

1.  $A = CDC^{-1} \rightsquigarrow A^2 = CD^2C^{-1}$

2.  $A^{-1} = CD^{-1}C^{-1}$  as long as 0 not an eigenval.

3.  $A^2 = CDC^{-1} \rightsquigarrow A = "CD^{1/2}C^{-1}"$

4.  $A = CDC^{-1}$   $B = PAP^{-1}$   
 $\rightsquigarrow B = PCDC^{-1}P^{-1}$   
 $= QDQ^{-1}$   $Q = PC$

## Distinct Eigenvalues

Fact. If  $A$  has  $n$  distinct eigenvalues, then  $A$  is diagonalizable.

Why?

$n$  distinct eigenvalues

$\leadsto n$  lin ind  
eigenvectors.

## Non-Distinct Eigenvalues

Theorem. Suppose

- $A = n \times n$ , has eigenvalues  $\lambda_1, \dots, \lambda_k$
- $a_i =$  algebraic multiplicity of  $\lambda_i$
- $d_i =$  dimension of  $\lambda_i$  eigenspace ("geometric multiplicity")

Then

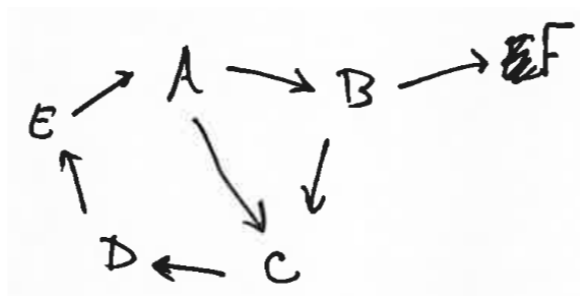
1.  $d_i \leq a_i$  for all  $i$  *actually,  $1 \leq d_i \leq a_i$*
2.  $A$  is diagonalizable  $\Leftrightarrow \sum d_i = n$   
*( $\Leftrightarrow \sum a_i = n$  and  $d_i = a_i$  for all  $i$ )*

## Example

Use your diagonalization of  $\begin{pmatrix} 0 & 1 \\ 1 & 1 \end{pmatrix}$  to find a formula for the  $n^{\text{th}}$  Fibonacci number.

# Application: Social Networks

Consider the social network below.



- We want to find **communities**, say, a group of people so there is a direct path connecting any two.
- Make a matrix,  $M$ , whose  $ij$ -entry is the number of arrows from  $i$  to  $j$ .
- Then the  $ij$  entry of  $M^2$  is the number of paths of length 2 to  $i$  to  $j$ . Why?
- Similar for  $M^3$ , etc.
- So the  $ij$  entry of  $M + M^2 + \dots + M^k$  is the number of paths of length at most  $k$ . We look for positive minors.

The leading eigenvalue is a measure of how connected the network is.



## Application: Business

Say your car rental company has 3 locations. Make a matrix  $M$  whose  $ij$  entry is the probability that a car at location  $i$  ends at location  $j$ . For example,

$$M = \begin{pmatrix} .3 & .4 & .5 \\ .3 & .4 & .3 \\ .4 & .2 & .2 \end{pmatrix}$$

Note the columns sum to 1. The eigenvector with eigenvalue 1 is the steady-state. Any other vector gets pulled to this state. Applying powers of  $M$  gives the state after some number of iterations.