

Announcements: Nov 6

- WebWork 5.1, 5.2 due Wednesday
- Quiz 5.1, 5.2 Friday
- Midterm 3 on Friday Nov 17
- Upcoming Office Hours
 - ▶ Me: Monday 1-2 and Wednesday 2-3, Skiles 234
 - ▶ Bharat: Tuesday 1:45-2:45, Skiles 230
 - ▶ Qianli: Wednesday 1-2, Clough 280
 - ▶ Arjun: Wednesday, 2:30-3:30, Skiles 230
 - ▶ Kemi: Thursday 9:30-10:30, Skiles 230
 - ▶ Martin: Friday 2-3, Skiles 230

Other help:

- ▶ Math Lab, Clough 280, Mon - Thu 12-6
- ▶ Tutoring: <http://www.successprograms.gatech.edu/tutoring>
- ▶ Group Tutoring
 - ▶ Sun Nov 12, 2-4, Clough 102
 - ▶ Sun Nov 19, 2-4, Clough 102

Orientation

Last week:

- Eigenvectors & eigenvalues
- (Similar matrices)

Today:

- Diagonalization
- Using diagonalization to take powers
- Algebraic versus geometric dimension

Eigenvectors and Eigenvalues

If A is an $n \times n$ matrix and there is a $v \neq 0$ in \mathbb{R}^n and λ in \mathbb{R} so

$$Av = \lambda v$$

then v is called an **eigenvector** for A , and λ is the corresponding **eigenvalue**.

Let A be an $n \times n$ matrix. The set of eigenvectors for a given eigenvalue λ of A (plus the zero vector) is a subspace of \mathbb{R}^n called the λ -**eigenspace** of A .

Fact. A invertible $\Leftrightarrow 0$ is not an eigenvalue of A

Similar matrices

Two $n \times n$ matrices A and B are **similar** if there is a matrix C so that

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

A is doing in the C -basis what B does in the regular basis.

Fact. If A and B similar, they have the same characteristic polynomial.

Similar matrices

Example

Similar: $A = CBC^{-1}$

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

A is doing in the C -basis what B does in the regular basis.

▶ Demo

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

Powers of diagonal matrices

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

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

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

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?

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

$$\begin{pmatrix} 1 & 2 \\ -1 & 4 \end{pmatrix} = \begin{pmatrix} 2 & 1 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} 2 & 0 \\ 0 & 3 \end{pmatrix} \begin{pmatrix} 2 & 1 \\ 1 & 1 \end{pmatrix}^{-1} \quad \text{“diagonalization”}$$

So...

Diagonalization

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

$$A = CDC^{-1} \quad D = \text{diagonal}$$

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

Diagonalization

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

In this case

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

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

Why?

Example

Diagonalize if possible.

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

Example

Diagonalize if possible.

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

Example

Diagonalize if possible.

$$\begin{pmatrix} 3/4 & 1/4 \\ 1/4 & 3/4 \end{pmatrix}$$

▶ Demo

More Examples

Diagonalize if possible.

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

Poll

Poll

Which are diagonalizable?

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

Distinct Eigenvalues

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

Why?

Non-Distinct Eigenvalues

Theorem. Suppose

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

Then

1. $d_i \leq a_i$ for all i
2. A is diagonalizable $\Leftrightarrow \sum d_i = n$
 $\Leftrightarrow \sum a_i = n$ and $d_i = a_i$ for all i

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.

Why is this similar/same as Google?

Summary

- A is diagonalizable if it is similar to a diagonal matrix
- If $A = CDC^{-1}$ then $A^k = CD^kC^{-1}$
- A is diagonalizable $\Leftrightarrow A$ has n linearly independent eigenvectors \Leftrightarrow the sum of the geometric dimensions of the eigenspaces is n
- If A has n distinct eigenvalues it is diagonalizable