

Section 5.2 : The Characteristic Equation

Chapter 5 : Eigenvalues and Eigenvectors
Math 1554 Linear Algebra

Topics

We will cover these topics in this section.

1. The characteristic polynomial of a matrix
2. Algebraic and geometric multiplicity of eigenvalues
3. Similar matrices

Objectives

For the topics covered in this section, students are expected to be able to do the following.

1. Construct the characteristic polynomial of a matrix and use it to identify eigenvalues and their multiplicities.
2. Characterize the long-term behaviour of dynamical systems using eigenvalue decompositions.

Section 5.2

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Exam 2 on
Wed @ 6:30pm
Cover up to [85.2]

How to find the λ 's.

The Characteristic Polynomial

Recall:

λ is an eigenvalue of $A \Leftrightarrow (A - \lambda I)$ is not invertible.

Therefore, to calculate the eigenvalues of A , we can solve

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

The quantity $\det(A - \lambda I)$ is the characteristic polynomial of A .

The quantity $\det(A - \lambda I) = 0$ is the characteristic equation of A .

The roots of the characteristic polynomial are the eigenvalues of A .

$$\text{Set } p(\lambda) = 0$$

$$-(a-b) = (b-a)$$

Example

The characteristic polynomial of $A = \begin{pmatrix} 5 & 2 \\ 2 & 1 \end{pmatrix}$ is:

$$p(\lambda) = \det(A - \lambda I) = \det \begin{bmatrix} 5-\lambda & 2 \\ 2 & 1-\lambda \end{bmatrix}$$

So the eigenvalues of A are:

$$= (5-\lambda)(1-\lambda) - (2)(2)$$

Why?

roots of $p(\lambda)$ are

$$-\frac{-b \pm \sqrt{b^2 - 4ac}}{2a} = \lambda_1, \lambda_2$$

$$\lambda = \frac{6 \pm \sqrt{36-4}}{2}$$

$$\lambda = 3 \pm \frac{\sqrt{32}}{2} = \frac{\sqrt{32}}{2} = \pm \sqrt{8} = \pm 2\sqrt{2}$$

$$\lambda = 3 \pm \frac{4\sqrt{2}}{2}$$

$$\lambda = 3 \pm 2\sqrt{2}$$

$$\begin{aligned} &= (\lambda-5)(\lambda-1) - 4 \\ &= \lambda^2 - 6\lambda + 5 - 4 \\ &= \lambda^2 - 6\lambda + 1 \end{aligned}$$

$$\uparrow p(\lambda) \text{ the char poly of } A.$$

$$\text{Roots } p(\lambda) = 0 \text{ solve for } \lambda$$

why this?

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

$A - \lambda I$ not invertible.
($\lambda \text{ EFT}$)

$$\Leftrightarrow (A - \lambda I)\vec{x} = \vec{0}$$

non-zero soln $\vec{x} \neq \vec{0}$.

$$\Leftrightarrow A\vec{x} = \lambda\vec{x} \quad (\vec{x} \neq \vec{0})$$

$$A = \begin{bmatrix} 5 & 2 \\ 2 & 1 \end{bmatrix} \quad p(\lambda) = \lambda^2 - 6\lambda + 1$$

Characteristic Polynomial of 2x2 Matrices

Express the characteristic equation of

$$M = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$$

in terms of its determinant. What is the equation when M is singular?

$$\begin{aligned} p(\lambda) &= \det \begin{bmatrix} a-\lambda & b \\ c & d-\lambda \end{bmatrix} \\ &= (a-\lambda)(d-\lambda) - bc = (\lambda-a)(\lambda-d) - bc \\ &= \lambda^2 - a\lambda - d\lambda + ad - bc \\ &= \lambda^2 - (a+d)\lambda + (ad - bc) \end{aligned}$$

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$$A = \begin{bmatrix} 5 & 2 \\ 2 & 1 \end{bmatrix} \quad p(\lambda) = \lambda^2 - [6]\lambda + [1]$$

let A

$\operatorname{tr}(A)$

"The sum of diag entries"

Geometric Multiplicity

#free vars in $A - \lambda I$.

Definition

The geometric multiplicity of an eigenvalue λ is the dimension of $\operatorname{Null}(A - \lambda I)$.

1. Geometric multiplicity is always at least 1. It can be smaller than algebraic multiplicity.

2. Here is the basic example:

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

$\lambda = 0$ is the only eigenvalue. Its algebraic multiplicity is 2, but the geometric multiplicity is 1.

$$\lambda_1 = 0 \quad \lambda_2 = 0 \quad \text{alg mult is } 2$$

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geo mult is dim $\operatorname{Null}(A - \lambda I)$.

$$\text{def. } A - 0I = A = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} \quad \vec{x} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$$\operatorname{Null} A = \operatorname{span} \left\{ \begin{pmatrix} 0 \\ 0 \end{pmatrix} \right\} \quad \text{dim Null } A = 1$$

so 0-eigenspace has dim of 1

geo mult of $\lambda = 0$ is 1

#times the root appears in $p(\lambda) = 0$

Definition

The algebraic multiplicity of an eigenvalue is its multiplicity as a root of the characteristic polynomial.

Example

Compute the algebraic multiplicities of the eigenvalues for the matrix

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

$\lambda_1 = 1, \lambda_2 = 0, \lambda_3 = -1, \lambda_4 = 0$

deg mult 1

deg mult 2

deg mult 3

deg mult 4

$$\begin{aligned} \det(A - \lambda I) &= \det \begin{bmatrix} 1-\lambda & 1 & 1 & 1 \\ 0 & 2-\lambda & 0 & 0 \\ 0 & 0 & -1-\lambda & 0 \\ 0 & 0 & 0 & 2-\lambda \end{bmatrix} \\ &= (1-\lambda)(-\lambda)(-1-\lambda)(-\lambda) \\ &= \lambda^2(\lambda-1)(\lambda+1) \\ &= (\lambda-0)^2(\lambda-1)(\lambda+1)^2 \end{aligned}$$

$$\begin{aligned} \lambda^2(\lambda^2-1) &?? \\ &= \lambda^4 - \lambda^2 = 0 \end{aligned}$$

FACT. A $n \times n$ matrix \forall

$$p(\lambda) = (-1)^n \lambda^n + \text{term } \lambda^{n-1}$$

$$+ \dots + \text{term}$$

deg $n-1$ term

deg $n-2$ term

deg $n-3$ term

deg $n-4$ term

$\lambda = 3$ alg mult 4 & geo mult 1. 3 instead.

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

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

$$\Rightarrow A - 3I = \begin{pmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \end{pmatrix}$$

$$\text{rank: } A - 3I = \begin{pmatrix} 0 & 1 & 2 & 2 \\ 0 & 0 & 1 & 2 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 \end{pmatrix}$$

Count free vars

$$\vec{x} = \begin{pmatrix} 1 \\ 0 \\ 0 \\ 0 \end{pmatrix}$$

basis for $\lambda = 3$ eigenspace is $\left\{ \begin{pmatrix} 1 \\ 0 \\ 0 \\ 0 \end{pmatrix} \right\}$



so geo mult is 1.

You try it!

Construct a matrix A w/
specified λ 's
specified alg mult.
specified geo mult.

$$\begin{aligned} \text{Ex: } & \lambda_1 = 1 \text{ alg 2 geo 1} \\ & \lambda_2 = 5 \text{ alg 3 geo 2} \end{aligned}$$

Recall: Long-Term Behavior of Markov Chains

Recall:

- We often want to know what happens to a Markov Chain
- $$\vec{x}_{k+1} = P\vec{x}_k, \quad k = 0, 1, 2, \dots$$
- as $k \rightarrow \infty$.
- If P is regular, then there is a _____

Now lets ask:

- If we don't know whether P is regular, what else might we do to describe the long-term behavior of the system?
- What can eigenvalues tell us about the behavior of these systems?

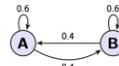
Example: Eigenvalues and Markov Chains

Note: the textbook has a similar example that you can review.

Consider the Markov Chain:

$$\vec{x}_{k+1} = P\vec{x}_k = \begin{pmatrix} 0.6 & 0.4 \\ 0.4 & 0.6 \end{pmatrix} \vec{x}_k, \quad k = 0, 1, 2, 3, \dots, \quad \vec{x}_0 = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$

This system can be represented schematically with two nodes, A and B:



What are the eigenvalues of P ?

What are the corresponding eigenvectors of P ?

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$$x = \begin{pmatrix} 3/2 \\ 1 \end{pmatrix}, \quad Px = \vec{v}$$

$$\det(AA^{-1}) = \det I = 1$$

$$\det A \det A^{-1} = 1$$

$$\det A^{-1} = \frac{1}{\det A}$$

Use the eigenvalues and eigenvectors of P to analyze the long-term behaviour of the system. In other words, determine what \vec{x}_k tends to as $k \rightarrow \infty$.

$A, B \text{ } n \times n$.

$$\det(AB) = \det A * \det B$$

$$\begin{aligned} y &= 0 \\ x \in \mathbb{R}^n &\text{ s.t. } \\ Ax &\neq A\vec{0} \quad \forall \vec{x} \neq \vec{0} \end{aligned} \quad \left\{ \begin{array}{l} A(\vec{v} + \vec{w}) = A\vec{v} + A\vec{w} \\ = \vec{v} + \vec{w} = 2\vec{v} + \vec{w} \\ \left(\begin{array}{c|c} 1 & 1 \\ 1 & 1 \end{array} \right) \neq \left(\begin{array}{c|c} 2 & 1 \\ 1 & 1 \end{array} \right) \\ = \left(\begin{array}{c} 4 \\ 4 \end{array} \right) = 2\vec{v} \end{array} \right.$$

$$\left(\begin{array}{c|c} 1 & 1 \\ 1 & 1 \end{array} \right) \vec{x} = \left(\begin{array}{c} 2 \\ 2 \end{array} \right)$$

$$\left(\begin{array}{c} 2 \\ 0 \end{array} \right), \left(\begin{array}{c} 1 \\ 1 \end{array} \right), \left(\begin{array}{c} -1 \\ 3 \end{array} \right) \quad \checkmark$$

Similar Matrices

Definition

Two $n \times n$ matrices A and B are **similar** if there is a matrix P so that $A = PBP^{-1}$.

Theorem

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

If time permits, we will explain or prove this theorem in lecture. Note:

- Our textbook introduces similar matrices in Section 5.2, but doesn't have exercises on this concept until 5.3.
- Two matrices, A and B , do not need to be similar to have the same eigenvalues. For example,

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

Additional Examples (if time permits)

1. True or false.

True) a) If A is similar to the identity matrix, then A is equal to the identity matrix.

False) b) A row replacement operation on a matrix does not change its eigenvalues.

2. For what values of k does the matrix have one real eigenvalue with algebraic multiplicity 2?

$$\begin{pmatrix} -3 & k \\ 2 & -6 \end{pmatrix}$$

$$P(\lambda) = \det \begin{pmatrix} -3 - \lambda & k \\ 2 & -6 - \lambda \end{pmatrix}$$

$$= (-3 - \lambda)(-6 - \lambda) - 2k$$

$$= (\lambda + 3)(\lambda + 6) - 2k$$

$$= \lambda^2 + 9\lambda + 18 - 2k$$

$$= (\lambda + 9)^2$$

$$18 - 2k = \left(-\frac{9}{2}\right)^2$$

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$$I = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

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

$$PA(\lambda) = \lambda^2 - 2\lambda + 1$$

$$PI(\lambda) = \lambda^2 - 2\lambda + 1$$

$$PA(\lambda) = \begin{pmatrix} 1-\lambda & 1 \\ 0 & 1-\lambda \end{pmatrix} = (1-\lambda)^2 = \lambda^2 - 2\lambda + 1$$

IF $A \sim B$
Satisfy all
these.
Then

- (i.e. A, B have same char poly)
Then
* Same λ 's eigenvalues
* Same det
* Same trace.
* Same alg for each λ .

$\Rightarrow A, B$
are similar?

$$A \text{ is similar to } I$$

$$A = PIP^{-1} \Rightarrow A = I$$

$$PIP^{-1} = P P^{-1} = I \quad \checkmark$$



$$x = \begin{pmatrix} 3/2 \\ 1 \end{pmatrix}, \quad P\vec{x} = \vec{p}$$

$$\det(AA^{-1}) = \det I = 1$$

$$\boxed{\det A \det A^{-1} = 1} \quad \boxed{\det A^{-1} = \frac{1}{\det A}}$$

$$\begin{aligned} y &= 0 \\ x &\in \mathbb{R}^n \text{ s.t.} \\ Ax &\neq A\vec{0} \quad \forall \vec{x} \neq \vec{0} \end{aligned}$$
$$A(v+w) = Av + Aw$$
$$= b+b = 2b \neq b.$$
$$\begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} 3 \\ 1 \end{pmatrix} \neq \begin{pmatrix} 2 \\ 2 \end{pmatrix},$$
$$= \begin{pmatrix} 4 \\ 4 \end{pmatrix} = 2b$$

A, B $n \times n$.

$$\begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix} \vec{x} = \begin{pmatrix} 2 \\ 2 \end{pmatrix}$$

$$\det(AB) = \det A \cdot \det B$$

$$\begin{pmatrix} 2 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 \\ 1 \end{pmatrix}, \begin{pmatrix} -1 \\ 3 \end{pmatrix} \quad \checkmark$$

5.2 Exercises

Find the characteristic polynomial and the eigenvalues of the matrices in Exercises 1–8.

$$1. \begin{bmatrix} 2 & 7 \\ 7 & 2 \end{bmatrix}$$

$$2. \begin{bmatrix} 5 & 3 \\ 3 & 5 \end{bmatrix}$$

$$3. \begin{bmatrix} 3 & -2 \\ 1 & -1 \end{bmatrix}$$

$$4. \begin{bmatrix} 4 & -3 \\ -4 & 2 \end{bmatrix}$$

$$5. \begin{bmatrix} 2 & 1 \\ -1 & 4 \end{bmatrix}$$

$$6. \begin{bmatrix} 1 & -4 \\ 4 & 6 \end{bmatrix}$$

$$7. \begin{bmatrix} 5 & 3 \\ -4 & 4 \end{bmatrix}$$

$$8. \begin{bmatrix} 7 & -2 \\ 2 & 3 \end{bmatrix}$$

Exercises 9–14 require techniques from Section 3.1. Find the characteristic polynomial of each matrix using expansion across a row or down a column. [Note: Finding the characteristic polynomial of a 3×3 matrix is not easy to do with just row operations, because the variable λ is involved.]

$$9. \begin{bmatrix} 1 & 0 & -1 \\ 2 & 3 & -1 \\ 0 & 6 & 0 \end{bmatrix}$$

$$10. \begin{bmatrix} 0 & 3 & 1 \\ 3 & 0 & 2 \\ 1 & 2 & 0 \end{bmatrix}$$

$$11. \begin{bmatrix} 4 & 0 & 0 \\ 5 & 3 & 2 \\ -2 & 0 & 2 \end{bmatrix}$$

$$12. \begin{bmatrix} 1 & 0 & 1 \\ -3 & 6 & 1 \\ 0 & 0 & 4 \end{bmatrix}$$

$$13. \begin{bmatrix} 6 & -2 & 0 \\ -2 & 9 & 0 \\ 5 & 8 & 3 \end{bmatrix}$$

$$14. \begin{bmatrix} 3 & -2 & 3 \\ 0 & -1 & 0 \\ 6 & 7 & -4 \end{bmatrix}$$

For the matrices in Exercises 15–17, list the eigenvalues, repeated according to their multiplicities.

$$15. \begin{bmatrix} 4 & -7 & 0 & 2 \\ 0 & 3 & -4 & 6 \\ 0 & 0 & 3 & -8 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

$$16. \begin{bmatrix} 5 & 0 & 0 & 0 \\ 8 & -4 & 0 & 0 \\ 0 & 7 & 1 & 0 \\ 1 & -5 & 2 & 1 \end{bmatrix}$$

$$17. \begin{bmatrix} 3 & 0 & 0 & 0 & 0 \\ -5 & 1 & 0 & 0 & 0 \\ 3 & 8 & 0 & 0 & 0 \\ 0 & -7 & 2 & 1 & 0 \\ -4 & 1 & 9 & -2 & 3 \end{bmatrix}$$

18. It can be shown that the algebraic multiplicity of an eigenvalue λ is always greater than or equal to the dimension of the eigenspace corresponding to λ . Find h in the matrix A below such that the eigenspace for $\lambda = 5$ is two-dimensional:

$$A = \begin{bmatrix} 5 & -2 & 6 & -1 \\ 0 & 3 & h & 0 \\ 0 & 0 & 5 & 4 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

19. Let A be an $n \times n$ matrix, and suppose A has n real eigenvalues, $\lambda_1, \dots, \lambda_n$, repeated according to multiplicities, so that $\det(A - \lambda I) = (\lambda_1 - \lambda)(\lambda_2 - \lambda) \cdots (\lambda_n - \lambda)$

In Exercises 21–30, A and B are $n \times n$ matrices. Mark each statement True or False (T/F). Justify each answer.

21. (T/F) If 0 is an eigenvalue of A , then A is invertible.

22. (T/F) The zero vector is in the eigenspace of A associated with an eigenvalue λ .

23. (T/F) The matrix A and its transpose, A^T , have different sets of eigenvalues.

24. (T/F) The matrices A and $B^{-1}AB$ have the same sets of eigenvalues for every invertible matrix B .

25. (T/F) If 2 is an eigenvalue of A , then $A - 2I$ is not invertible.

26. (T/F) If two matrices have the same set of eigenvalues, then they are similar.

27. (T/F) If $\lambda + 5$ is a factor of the characteristic polynomial of A , then 5 is an eigenvalue of A .

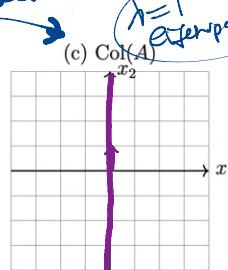
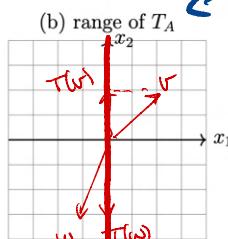
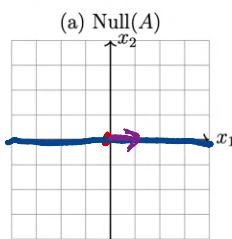
28. (T/F) The multiplicity of a root r of the characteristic equation of A is called the algebraic multiplicity of r as an eigenvalue of A .

29. (T/F) The eigenvalue of the $n \times n$ identity matrix is 1 with algebraic multiplicity n .

30. (T/F) The matrix A can have more than n eigenvalues.

Midterm 2 Lecture Review Activity, Math 1554

1. (3 points) T_A is the linear transform $x \rightarrow Ax$, $A \in \mathbb{R}^{2 \times 2}$, that projects points in \mathbb{R}^2 onto the x_2 -axis. Sketch the nullspace of A , the range of the transform, and the column space of A . How are the range and column space related to each other?



The range is what's come out

$$T(\vec{x}) = \vec{b}$$

But

$$T(\vec{x}) = A\vec{x}$$

$$= x_1v_1 + x_2v_2 = \vec{b}$$

META: $A = [T(e_1) \ T(e_2)] = \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} \sim \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} \quad \vec{x} = r \begin{pmatrix} 1 \\ 0 \end{pmatrix}$

$$T\left(\begin{pmatrix} 1 \\ 0 \end{pmatrix}\right) = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$$T\left(\begin{pmatrix} 0 \\ 1 \end{pmatrix}\right) = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$$\text{Null } A = \text{Span}\left\{\begin{pmatrix} 1 \\ 0 \end{pmatrix}\right\}$$

$$\text{Col } A = \text{Span}\left\{\begin{pmatrix} 0 \\ 1 \end{pmatrix}, \begin{pmatrix} 0 \\ 0 \end{pmatrix}\right\}$$

$$= \text{Span}\left\{\begin{pmatrix} 0 \\ 1 \end{pmatrix}\right\}$$

2. Indicate true if the statement is true, otherwise, indicate false.

for some $a \neq 4$

a) $S = \{\vec{x} \in \mathbb{R}^3 \mid x_1 = 4, x_2 = 16, x_3 = x_1x_2\}$ is a subspace for any $a \in \mathbb{R}$.

true false



b) If A is square and non-zero, and $A\vec{x} = A\vec{y}$ for some $\vec{x} \neq \vec{y}$, then $\det(A) \neq 0$.



① IF $0 \notin S$ then S is not a subspace

Check ① IF $\vec{0} \in S$ then $\vec{0} \in S$

② If $\vec{v} \in S$ and $c \in \mathbb{R}$ then $c\vec{v} \in S$.

If "for any" universal quantifier
If "for some" existential quantifier
quantified.

(a)

$$S = \left\{ \begin{pmatrix} 4 \\ 16 \\ 64 \end{pmatrix} \right\}$$

~~T/F~~ S is a 1-dimensional Subspace of \mathbb{R}^3

alt approach

$$Ax - Ay = \vec{0} \\ \Rightarrow A(x - y) = \vec{0}$$

$$x \neq y \text{ so} \\ x - y \neq \vec{0}$$

(b) If $x \neq y$ and $T(x) = T(y)$

then T is not 1-to-1

$\Rightarrow A$ is not inv. $\Rightarrow \det = 0$.

Q: What is a steady-state vector of a stochastic matrix P ?

Q₂: What is the long term trend of a Markov chain?

Q₃: Are they the same thing?

$$\begin{bmatrix} .2 \\ .8 \end{bmatrix}, \begin{bmatrix} .8 \\ .2 \end{bmatrix}, \begin{bmatrix} .2 \\ .8 \end{bmatrix}, \begin{bmatrix} .8 \\ .2 \end{bmatrix}, \dots$$

Markov Chain.

Ex. $P = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$ $x_0 = \begin{bmatrix} .2 \\ .8 \end{bmatrix}$

$$x_1 = \begin{bmatrix} .2 \\ .8 \end{bmatrix}, \quad x_1 = P x_0 = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} .2 \\ .8 \end{bmatrix} = \begin{bmatrix} .8 \\ .2 \end{bmatrix}$$

$$x_2 = P^2 x_0 = \begin{bmatrix} .2 \\ .8 \end{bmatrix}$$

$$x_3 = P^3 x_0 = \begin{bmatrix} .2 \\ .8 \end{bmatrix}$$

$$P x_0 = x_0$$

$$P - I = \begin{pmatrix} -1 & 1 \\ 1 & -1 \end{pmatrix} \sim \begin{pmatrix} 1 & -1 \\ 0 & 0 \end{pmatrix} \quad \vec{x} = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

$$x_0 = \begin{pmatrix} .5 \\ .5 \end{pmatrix}$$

So x_0 is
a steady-state

3. If possible, write down an example of a matrix or quantity with the given properties. If it is not possible to do so, write *not possible*.

(a) A is 2×2 , $\text{Col}A$ is spanned by the vector $\begin{pmatrix} 2 \\ 3 \end{pmatrix}$ and $\dim(\text{Null}(A)) = 1$. $A = \begin{pmatrix} 2 & 2 \\ 3 & 3 \end{pmatrix} \sim \begin{pmatrix} 2 & 2 \\ 0 & 0 \end{pmatrix}$

(b) A is 2×2 , $\text{Col}A$ is spanned by the vector $\begin{pmatrix} 2 \\ 3 \end{pmatrix}$ and $\dim(\text{Null}(A)) = 0$. $A = \begin{pmatrix} NP \end{pmatrix}$

(c) A is in RREF and $T_A : \mathbb{R}^3 \rightarrow \mathbb{R}^3$. The vectors u and v are a basis for the range of T .

$$u = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}, v = \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}, A = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix}$$

basis for $\text{Col}A$.

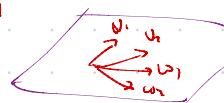
$$\text{Col}A = \text{span} \left\{ \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} \right\} \stackrel{?}{=} \text{span} \left\{ \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix} \right\}$$

yes.

In general $\text{Span}\{v_1, v_2\} = \text{Span}\{w_1, w_2\}$

$\Leftrightarrow v_1, v_2 \in \text{span}\{w_1, w_2\}$

AND $w_1, w_2 \in \text{span}\{v_1, v_2\}$



4. Indicate whether the situations are possible or impossible by filling in the appropriate circle.

$\lambda \neq \mu$ eigenvalues are different

possible impossible

4.i) Vectors \vec{u} and \vec{v} are eigenvectors of square matrix A , and $\vec{w} = \vec{u} + \vec{v}$ is also an eigenvector of A .



4.ii) $T_A = A\vec{x}$ is one-to-one, $\dim(\text{Col}(A)) = 4$, and $T_A : \mathbb{R}^3 \rightarrow \mathbb{R}^4$.



$$\begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \quad \lambda=1, \mu=-1$$

$$u = \begin{pmatrix} 1 \\ 1 \end{pmatrix}, v = \begin{pmatrix} -1 \\ 1 \end{pmatrix}$$

$$\begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} 1 \\ 1 \end{pmatrix} = \begin{pmatrix} 1 \\ 1 \end{pmatrix} \checkmark$$

$$\begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} -1 \\ 1 \end{pmatrix} = \begin{pmatrix} -1 \\ 1 \end{pmatrix} = \text{E} \begin{pmatrix} -1 \\ 1 \end{pmatrix}$$

$\lambda=1$
 $\mu=-1$

$$w = u + v = \begin{pmatrix} 0 \\ 2 \end{pmatrix} ? \quad \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} 0 \\ 2 \end{pmatrix} = \begin{pmatrix} 2 \\ 0 \end{pmatrix} \neq \text{E} \begin{pmatrix} 0 \\ 2 \end{pmatrix}$$

5. (2 points) Fill in the blanks.

(a) If A is a 6×4 matrix in RREF and $\text{rank}(A) = 4$, what is the rank of A^T ? 4

(b) $T_A = A\vec{x}$, where $A \in \mathbb{R}^{2 \times 2}$, is a linear transform that first rotates vectors in \mathbb{R}^2 clockwise by π radians about the origin, then scales their x -component by a factor of 3, then projects them onto the x_1 axis. What is the value of $\det(A)$? 0

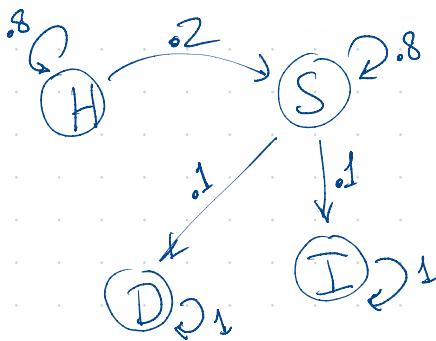
6. (3 points) A virus is spreading in a lake. Every week,

- 20% of the healthy fish get sick with the virus, while the other healthy fish remain healthy but could get sick at a later time.
- 10% of the sick fish recover and can no longer get sick from the virus, 80% of the sick fish remain sick, and 10% of the sick fish die.

Initially there are exactly 1000 fish in the lake.

a) What is the stochastic matrix, P , for this situation? Is P regular?

b) Write down any steady-state vector for the corresponding Markov-chain.



$$P = \begin{bmatrix} H & S & I & D \\ H & 0.8 & 0 & 0 & 0 \\ S & 0.2 & 0.9 & 0 & 0 \\ I & 0 & 0.1 & 1 & 0 \\ D & 0 & 0.1 & 0 & 1 \end{bmatrix}$$

Solve
 $(P - I)\vec{x} = 0$

$$P - I = \begin{bmatrix} -0.2 & 0 & 0 & 0 \\ 0.2 & -0.2 & 0 & 0 \\ 0 & 0.1 & 0 & 0 \\ 0 & 0.1 & 0 & 0 \end{bmatrix}$$

If not unique
→ not reg.

$$\sim \begin{bmatrix} -2 & 0 & 0 & 0 \\ 2 & -2 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

$$\sim \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

RHM
If reg → unique

$$\vec{x} = r \begin{pmatrix} 0 \\ 0 \\ 1 \\ 0 \end{pmatrix} + s \begin{pmatrix} 0 \\ 0 \\ 0 \\ 1 \end{pmatrix}$$

$$\begin{cases} s=1 \\ r=0 \end{cases} \quad \begin{pmatrix} 0 \\ 0 \\ 1 \\ 0 \end{pmatrix} \rightarrow \begin{pmatrix} 0 \\ 0 \\ 1/2 \\ 1/2 \end{pmatrix} \checkmark$$

$$\begin{cases} r=2 \\ s=3 \end{cases} \quad \begin{pmatrix} 0 \\ 0 \\ 2 \\ 3 \end{pmatrix} \rightarrow \begin{pmatrix} 0 \\ 0 \\ 4 \\ 6 \end{pmatrix} \checkmark$$

$$P = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \quad \begin{pmatrix} 0.5 \\ -0.5 \end{pmatrix} = \vec{q}$$

$$\vec{x} = s \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

6. (3 points) A virus is spreading in a lake. Every week,

- 20% of the healthy fish get sick with the virus, while the other healthy fish remain healthy but could get sick at a later time.
- 10% of the sick fish recover and can no longer get sick from the virus, 80% of the sick fish remain sick, and 10% of the sick fish die.

Initially there are exactly 1000 fish in the lake.

- What is the stochastic matrix, P , for this situation? Is P regular?
- Write down any steady-state vector for the corresponding Markov-chain.

Topics and Objectives

Section 5.3 : Diagonalization

Chapter 5 : Eigenvalues and Eigenvectors

Math 1554 Linear Algebra

Motivation: it can be useful to take large powers of matrices, for example A^k , for large k .

But: multiplying two $n \times n$ matrices requires roughly n^3 computations. Is there a more efficient way to compute A^k ?

Topics

1. Diagonal, similar, and diagonalizable matrices
2. Diagonalizing matrices

Learning Objectives

For the topics covered in this section, students are expected to be able to do the following.

1. Determine whether a matrix can be diagonalized, and if possible diagonalize a square matrix.
2. Apply diagonalization to compute matrix powers.

Section 5.3 : Diagonalization

Chapter 5 : Eigenvalues and Eigenvectors

Math 1554 Linear Algebra

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Topics and Objectives

Topics

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1. Determine whether a matrix can be diagonalized, and if possible diagonalize a square matrix.
2. Apply diagonalization to compute matrix powers.

Course Schedule

Cancellations due to inclement weather will likely result in cancelling review lectures and possibly moving through course material.

Week Dates	Mon Lecture	Tue Studio	Wed Lecture	Thu Studio	Fri Lecture
1 8/21 - 8/25	1.1	WS1.1	1.2	WS1.2	1.3
2 8/28 - 9/1	1.4	WS1.3, 1.4	1.5	WS1.5	1.7
3 9/4 - 9/8	Break	WS1.7	1.8	WS1.8	1.9
4 9/11 - 9/15	2.1	WS1.9, 2.1	Exam 1, Review	Cancelled	2.2
5 9/18 - 9/22	2.3, 2.4	WS2.2, 2.3	2.5	WS2.4, 2.5	2.8
6 9/25 - 9/29	2.9	WS2.8, 2.9	3.1, 3.2	WS3.1, 3.2	3.3
7 10/2 - 10/6	4.7	WS3.4, 4.7	5.1, 5.2	WS4.1, 5.2	5.5
8 10/9 - 10/13	Break	Break	Exam 2, Review	Cancelled	5.3
9 10/16 - 10/20	5.3	WS5.3	5.5	WS5.5	6.1
10 10/23 - 10/27	6.1, 6.2	WS6.1	6.2	WS6.2	6.3
11 10/30 - 11/3	6.4	WS6.3, 6.4	6.4, 6.5	WS6.4, 6.5	6.5
12 11/6 - 11/10	6.6	WS6.5, 6.6	Exam 3, Review	Cancelled	PageRank
13 11/13 - 11/17	7.1	WS7.pageRank	7.2	WS7.1, 7.2	7.3
14 11/20 - 11/24	7.3, 7.4	WS7.2, 7.3	Break	Break	Break
15 11/27 - 12/1	7.4	WS7.3, 7.4	7.4	WS7.4	7.4
16 12/4 - 12/8	Last lecture	Last Studio	Reading Period		
17 12/11 - 12/15	Final Exam: MATH 1554 Common Final Exam Tuesday, December 12th at 6pm				

Diagonal Matrices

A matrix is **diagonal** if the only non-zero elements, if any, are on the main diagonal.

The following are all diagonal matrices.

$$\begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix}, \quad [2], \quad I_n, \quad \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$$

We'll only be working with diagonal square matrices in this course.

Powers of Diagonal Matrices

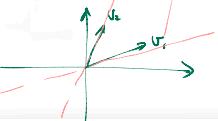
If A is diagonal, then A^k is easy to compute. For example,

$$A = \begin{pmatrix} 3 & 0 \\ 0 & 0.5 \end{pmatrix}$$

$$A^2 = \begin{pmatrix} 9 & 0 \\ 0 & 0.25 \end{pmatrix}$$

$$A^k = \begin{pmatrix} 3^k & 0 \\ 0 & (0.5)^k \end{pmatrix}$$

But what if A is not diagonal?



A diagonal matrix A has the property that the eigenvectors are the diagonal entries. The eigenvalues are the diagonal entries.

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

$$Ae_1 = \begin{pmatrix} n \\ 0 \end{pmatrix} = ne_1$$

$$(Ae_1 = ne_1)$$

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Diagonalization

Suppose $A \in \mathbb{R}^{n \times n}$. We say that A is **diagonalizable** if it is similar to a diagonal matrix, D . That is, we can write

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

(what are these)

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

Diagonalization

Theorem

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

Note: the symbol \Leftrightarrow means "if and only if".

Also note that $A = PDP^{-1}$ if and only if

$$A = [\vec{v}_1 \ \vec{v}_2 \ \dots \ \vec{v}_n] \begin{bmatrix} \lambda_1 & & & \\ & \lambda_2 & & \\ & & \ddots & \\ & & & \lambda_n \end{bmatrix} [\vec{v}_1 \ \vec{v}_2 \ \dots \ \vec{v}_n]^{-1}$$

where $\vec{v}_1, \dots, \vec{v}_n$ are linearly independent eigenvectors, and $\lambda_1, \dots, \lambda_n$ are the corresponding eigenvalues (in order).

$\lambda = 1 \text{ alg 2}$

$$A = \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix}$$

geo mult 1

1

$$A - I = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$$

geo 1

$$\vec{x} = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

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$$D = \begin{bmatrix} \lambda_1 & 0 & 0 \\ 0 & \lambda_2 & 0 \\ 0 & 0 & \lambda_3 \end{bmatrix}$$

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$$P = \begin{bmatrix} | & | & | \\ \vec{v}_1 & \vec{v}_2 & \vec{v}_3 \\ | & | & | \end{bmatrix}$$

What could go wrong

$$A\vec{v}_i = \lambda_i \vec{v}_i$$

No real eigenvalues

Check $\begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix} = A$

$$p(\lambda) = |A - \lambda I| = \begin{vmatrix} -\lambda & -1 \\ 1 & -\lambda \end{vmatrix}$$

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

$$= \lambda^2 + 1$$

compute $\lambda = \pm i$

$$= (\lambda - i)(\lambda + i) = 0$$

Distinct Eigenvalues

Theorem

If A is $n \times n$ and has n distinct eigenvalues, then A is diagonalizable.

Why does this theorem hold?

Is it necessary for an $n \times n$ matrix to have n distinct eigenvalues for it to be diagonalizable?

Non-Distinct Eigenvalues

Theorem. Suppose

- A is $n \times n$
- A has distinct eigenvalues $\lambda_1, \dots, \lambda_k$, $k \leq n$
- 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 $\Leftrightarrow \sum d_i = n \Leftrightarrow d_i = a_i$ for all i
3. A is diagonalizable \Leftrightarrow the eigenvectors, for all eigenvalues, together form a basis for \mathbb{R}^n .

Example 1

Shear

Diagonalize if possible.

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

Step 1: Find λ_1 & λ_2 Step 2: Find a basis of eigenvectors for each eigenspaceStep 3: Construct $P = \begin{pmatrix} \vec{v}_1 & \vec{v}_2 \end{pmatrix}$

$$D = \begin{pmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{pmatrix}$$

Step 4 (optional): Check $A = PDP^{-1}$?

Example 2

Diagonalize if possible.

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

$$\text{Step 1: } p(\lambda) = \begin{vmatrix} 2-\lambda & 6 \\ 0 & -1-\lambda \end{vmatrix}$$

$$= (2-\lambda)(-1-\lambda) - 0$$

$$= (\lambda-2)(\lambda+1) = 0$$

$$\lambda = 2, -1$$

$$\text{Step 2: } \lambda = 2 \quad A - 2I = \begin{bmatrix} 0 & 6 \\ 0 & -3 \end{bmatrix} \sim \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \quad \vec{x} = c \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

basis is $\left\{ \begin{bmatrix} 1 \\ 0 \end{bmatrix} \right\}$

$$\lambda = -1 \quad A + I = \begin{bmatrix} 3 & 6 \\ 0 & 0 \end{bmatrix} \sim \begin{bmatrix} 1 & 2 \\ 0 & 0 \end{bmatrix} \quad \vec{x} = r \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

basis is $\left\{ \begin{bmatrix} 1 \\ 1 \end{bmatrix} \right\}$

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

???

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

$$A = Q \overset{\curvearrowleft}{D} Q^{-1}$$

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

Example 2

Diagonalize if possible.

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

$\lambda=3$ only eigenvalue.

Step 1: $p(\lambda) = (\lambda-3)^2 = 0$ alg 2

Step 2: $\lambda=3 \quad A-3I = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} \quad \vec{x} = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$

geo mult is 1.

blk only one

free var

$$\text{so } \dim \text{Null}(A-3I) = 1.$$

(dim of eigenspace
= geo mult)

NOT invertible

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

IF $A = P \Lambda P^{-1}$
then $A = I$

Example 3

$$\text{deg mult? } p(\lambda) = -(\lambda-1)(\lambda-2)(\lambda-3)$$

The eigenvalues of A are $\lambda = 3, 1$. If possible, construct P and D such that $AP = PD$.

$$A = \begin{pmatrix} 7 & 4 & 16 \\ 2 & 5 & 8 \\ -2 & -2 & -5 \end{pmatrix}$$

$$\lambda_1=1 \quad A - I = \begin{pmatrix} 6 & 4 & 16 \\ 2 & 4 & 8 \\ -2 & -2 & -6 \end{pmatrix}$$

a basis for λ_1 -eigenspace

$$\sim \left[\begin{array}{ccc|c} 1 & 1 & 3 & 0 \\ 2 & 4 & 8 & 0 \\ 0 & 4 & 16 & 0 \end{array} \right] \sim \left[\begin{array}{ccc|c} 1 & 1 & 3 & 0 \\ 0 & 2 & 2 & 0 \\ 0 & 2 & 2 & 0 \end{array} \right] \sim \left[\begin{array}{ccc|c} 1 & 1 & 3 & 0 \\ 0 & 1 & 1 & 0 \\ 0 & 0 & 0 & 0 \end{array} \right]$$

see mathtut-1.

$$\sim \left[\begin{array}{ccc|c} 1 & 1 & 3 & 0 \\ 0 & 1 & 1 & 0 \\ 0 & 0 & 0 & 0 \end{array} \right] \sim \left[\begin{array}{ccc|c} 1 & 0 & 2 & 0 \\ 0 & 1 & 1 & 0 \\ 0 & 0 & 0 & 0 \end{array} \right] \quad \vec{x} = \begin{pmatrix} -2 \\ -1 \\ 1 \end{pmatrix}$$

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Additional Example (if time permits)

Note that

$$\vec{x}_k = \begin{bmatrix} 0 \\ 1 \\ \vdots \\ k \end{bmatrix}, \quad \vec{x}_0 = \begin{bmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}, \quad k = 1, 2, 3, \dots$$

generates a well-known sequence of numbers.

Use a diagonalization to find a matrix equation that gives the n^{th} number in this sequence.

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

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$$\lambda_2=3 \quad A - 3I = \begin{pmatrix} 4 & 4 & 16 \\ 2 & 2 & 8 \\ -2 & -2 & -8 \end{pmatrix}$$

$$\sim \left[\begin{array}{ccc|c} 1 & 1 & 4 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{array} \right] \quad \vec{x} = 1 \begin{pmatrix} -1 \\ 0 \\ 0 \end{pmatrix} + 3 \begin{pmatrix} -4 \\ 0 \\ 0 \end{pmatrix}$$

a basis for λ_2 -eigenspace is

$$\left\{ \begin{pmatrix} -1 \\ 0 \\ 0 \end{pmatrix}, \begin{pmatrix} -4 \\ 0 \\ 0 \end{pmatrix} \right\}$$

$$P = \begin{pmatrix} 1 & -1 & -4 \\ -1 & 1 & 0 \\ 1 & 0 & 1 \end{pmatrix}$$

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

$$A = \begin{pmatrix} * & * & * \\ * & * & * \\ * & * & * \end{pmatrix} \quad \text{Stochastic}$$

$$\begin{aligned} \lambda_1 &= 1 & v_1 &= \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} \\ \lambda_2 &= .5 & v_2 &= \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix} \\ \lambda_3 &= 0 & v_3 &= \begin{pmatrix} -1 \\ 1 \\ 2 \end{pmatrix} \end{aligned} \quad \left. \right\}$$

$$A^k \vec{x}_0 = \vec{x}_n$$

$$A^k \stackrel{?}{=} (PDP^{-1})^k$$

$$D^n = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 3 & 0 \\ 0 & 0 & 0 \end{pmatrix} = P D P^{-1} \cancel{P} \cancel{D} \cancel{P} \cancel{P} \cdots \cancel{P} \cancel{D} P^{-1}$$

$$\vec{x} \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix} \xrightarrow{k+1} = P D^k P^{-1} \quad \text{easy!}$$

Additional Example (if time permits)

Note that

$$\vec{x}_k = \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix} \vec{x}_{k-1}, \quad \vec{x}_0 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}, \quad k = 1, 2, 3, \dots$$

generates a well-known sequence of numbers.

Use a diagonalization to find a matrix equation that gives the n^{th} number in this sequence.

$$F = \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix} \quad F^k x_0 = x_n$$

$$x_0, x_1, x_2, \dots$$

$$(1), (1), (2), (3), (5), (8), (13), (21)$$

$$\begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix} \begin{pmatrix} 1 \\ 1 \end{pmatrix} = \begin{pmatrix} 1 \\ 2 \end{pmatrix}$$

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

$$\begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix} \begin{pmatrix} 2 \\ 3 \end{pmatrix} = \begin{pmatrix} 3 \\ 5 \end{pmatrix}$$

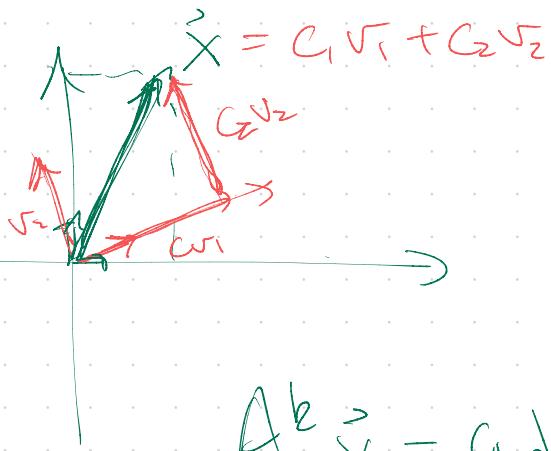
$$\begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix} \begin{pmatrix} 3 \\ 5 \end{pmatrix} = \begin{pmatrix} 5 \\ 8 \end{pmatrix}$$

$$1, 1, 2, 3, 5, 8, 13, 21, \dots$$

↑

10th value?

$$F^k \begin{pmatrix} 1 \\ 1 \end{pmatrix} = \begin{pmatrix} * \\ * \end{pmatrix}$$



$$A^k \vec{x} = C_1 \lambda_1^k \vec{v}_1 + C_2 \lambda_2^k \vec{v}_2$$

THEOREM 5

The Diagonalization Theorem

An $n \times n$ matrix A is diagonalizable if and only if A has n linearly independent eigenvectors.

In fact, $A = PDP^{-1}$, with D a diagonal matrix, if and only if the columns of P are n linearly independent eigenvectors of A . In this case, the diagonal entries of D are eigenvalues of A that correspond, respectively, to the eigenvectors in P .

EXAMPLE 4

Diagonalize the following matrix, if possible.

$$A = \begin{bmatrix} 2 & 4 & 3 \\ -4 & -6 & -3 \\ 3 & 3 & 1 \end{bmatrix}$$

THEOREM 6

An $n \times n$ matrix with n distinct eigenvalues is diagonalizable.

THEOREM 7

Let A be an $n \times n$ matrix whose distinct eigenvalues are $\lambda_1, \dots, \lambda_p$.

- For $1 \leq k \leq p$, the dimension of the eigenspace for λ_k is less than or equal to the multiplicity of the eigenvalue λ_k .
- The matrix A is diagonalizable if and only if the sum of the dimensions of the eigenspaces equals n , and this happens if and only if (i) the characteristic polynomial factors completely into linear factors and (ii) the dimension of the eigenspace for each λ_k equals the multiplicity of λ_k .
- If A is diagonalizable and B_k is a basis for the eigenspace corresponding to λ_k for each k , then the total collection of vectors in the sets B_1, \dots, B_p forms an eigenvector basis for \mathbb{R}^n .

5.3 EXERCISES

In Exercises 1 and 2, let $A = PDP^{-1}$ and compute A^4 .

1. $P = \begin{bmatrix} 5 & 7 \\ 2 & 3 \end{bmatrix}, D = \begin{bmatrix} 2 & 0 \\ 0 & 1 \end{bmatrix}$

2. $P = \begin{bmatrix} 2 & -3 \\ -3 & 5 \end{bmatrix}, D = \begin{bmatrix} 1 & 0 \\ 0 & 1/2 \end{bmatrix}$

In Exercises 3 and 4, use the factorization $A = PDP^{-1}$ to compute A^k , where k represents an arbitrary positive integer.

3. $\begin{bmatrix} a & 0 \\ 3(a-b) & b \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 3 & 1 \end{bmatrix} \begin{bmatrix} a & 0 \\ 0 & b \end{bmatrix} \begin{bmatrix} 1 & 0 \\ -3 & 1 \end{bmatrix}$

4. $\begin{bmatrix} -2 & 12 \\ -1 & 5 \end{bmatrix} = \begin{bmatrix} 3 & 4 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 2 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} -1 & 4 \\ 1 & -3 \end{bmatrix}$

In Exercises 5 and 6, the matrix A is factored in the form PDP^{-1} . Use the Diagonalization Theorem to find the eigenvalues of A and a basis for each eigenspace.

5. $\begin{bmatrix} 2 & 2 & 1 \\ 1 & 3 & 1 \\ 1 & 2 & 2 \end{bmatrix} =$

$\begin{bmatrix} 1 & 1 & 2 \\ 1 & 0 & -1 \\ 1 & -1 & 0 \end{bmatrix} \begin{bmatrix} 5 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1/4 & 1/2 & 1/4 \\ 1/4 & 1/2 & -3/4 \\ 1/4 & -1/2 & 1/4 \end{bmatrix}$

6. $\begin{bmatrix} 4 & 0 & -2 \\ 2 & 5 & 4 \\ 0 & 0 & 5 \end{bmatrix} =$

$\begin{bmatrix} -2 & 0 & -1 \\ 0 & 1 & 2 \\ 1 & 0 & 0 \end{bmatrix} \begin{bmatrix} 5 & 0 & 0 \\ 0 & 5 & 0 \\ 0 & 0 & 4 \end{bmatrix} \begin{bmatrix} 0 & 0 & 1 \\ 2 & 1 & 4 \\ -1 & 0 & -2 \end{bmatrix}$

Diagonalize the matrices in Exercises 7–20, if possible. The eigenvalues for Exercises 11–16 are as follows: (11) $\lambda = 1, 2, 3$; (12) $\lambda = 2, 8$; (13) $\lambda = 5, 1$; (14) $\lambda = 5, 4$; (15) $\lambda = 3, 1$; (16) $\lambda = 2, 1$. For Exercise 18, one eigenvalue is $\lambda = 5$ and one eigenvector is $(-2, 1, 2)$.

7. $\begin{bmatrix} 1 & 0 \\ 6 & -1 \end{bmatrix}$

8. $\begin{bmatrix} 5 & 1 \\ 0 & 5 \end{bmatrix}$

9. $\begin{bmatrix} 3 & -1 \\ 1 & 5 \end{bmatrix}$

10. $\begin{bmatrix} 2 & 3 \\ 4 & 1 \end{bmatrix}$

11. $\begin{bmatrix} -1 & 4 & -2 \\ -3 & 4 & 0 \\ -3 & 1 & 3 \end{bmatrix}$

12. $\begin{bmatrix} 4 & 2 & 2 \\ 2 & 4 & 2 \\ 2 & 2 & 4 \end{bmatrix}$

13. $\begin{bmatrix} 2 & 2 & -1 \\ 1 & 3 & -1 \\ -1 & -2 & 2 \end{bmatrix}$

14. $\begin{bmatrix} 4 & 0 & -2 \\ 2 & 5 & 4 \\ 0 & 0 & 5 \end{bmatrix}$

15. $\begin{bmatrix} 7 & 4 & 16 \\ 2 & 5 & 8 \\ -2 & -2 & -5 \end{bmatrix}$

16. $\begin{bmatrix} 0 & -4 & -6 \\ -1 & 0 & -3 \\ 1 & 2 & 5 \end{bmatrix}$

17. $\begin{bmatrix} 4 & 0 & 0 \\ 1 & 4 & 0 \\ 0 & 0 & 5 \end{bmatrix}$

18. $\begin{bmatrix} -7 & -16 & 4 \\ 6 & 13 & -2 \\ 12 & 16 & 1 \end{bmatrix}$

19. $\begin{bmatrix} 5 & -3 & 0 & 9 \\ 0 & 3 & 1 & -2 \\ 0 & 0 & 2 & 0 \\ 0 & 0 & 0 & 2 \end{bmatrix}$

20. $\begin{bmatrix} 4 & 0 & 0 & 0 \\ 0 & 4 & 0 & 0 \\ 0 & 0 & 2 & 0 \\ 1 & 0 & 0 & 2 \end{bmatrix}$

In Exercises 21 and 22, A , B , P , and D are $n \times n$ matrices. Mark each statement True or False. Justify each answer. (Study Theorems 5 and 6 and the examples in this section carefully before you try these exercises.)

21. a. A is diagonalizable if $A = PDP^{-1}$ for some matrix D and some invertible matrix P .

b. If \mathbb{R}^n has a basis of eigenvectors of A , then A is diagonalizable.

c. A is diagonalizable if and only if A has n eigenvalues, counting multiplicities.

d. If A is diagonalizable, then A is invertible.

22. a. A is diagonalizable if A has n eigenvectors.

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

c. If $AP = PD$, with D diagonal, then the nonzero columns of P must be eigenvectors of A .

d. If A is invertible, then A is diagonalizable.

23. A is a 5×5 matrix with two eigenvalues. One eigenspace is three-dimensional, and the other eigenspace is two-dimensional. Is A diagonalizable? Why?

24. A is a 3×3 matrix with two eigenvalues. Each eigenspace is one-dimensional. Is A diagonalizable? Why?
25. A is a 4×4 matrix with three eigenvalues. One eigenspace is one-dimensional, and one of the other eigenspaces is two-dimensional. Is it possible that A is *not* diagonalizable? Justify your answer.
26. A is a 7×7 matrix with three eigenvalues. One eigenspace is two-dimensional, and one of the other eigenspaces is three-dimensional. Is it possible that A is *not* diagonalizable? Justify your answer.
27. Show that if A is both diagonalizable and invertible, then so is A^{-1} .
28. Show that if A has n linearly independent eigenvectors, then so does A^T . [Hint: Use the Diagonalization Theorem.]
29. A factorization $A = PDP^{-1}$ is not unique. Demonstrate this for the matrix A in Example 2. With $D_1 = \begin{bmatrix} 3 & 0 \\ 0 & 5 \end{bmatrix}$, use the information in Example 2 to find a matrix P_1 such that $A = P_1 D_1 P_1^{-1}$.
30. With A and D as in Example 2, find an invertible P_2 unequal to the P in Example 2, such that $A = P_2 DP_2^{-1}$.
31. Construct a nonzero 2×2 matrix that is invertible but not diagonalizable.
32. Construct a nondiagonal 2×2 matrix that is diagonalizable but not invertible.

[M] Diagonalize the matrices in Exercises 33–36. Use your matrix program's eigenvalue command to find the eigenvalues, and then compute bases for the eigenspaces as in Section 5.1.

33. $\begin{bmatrix} -6 & 4 & 0 & 9 \\ -3 & 0 & 1 & 6 \\ -1 & -2 & 1 & 0 \\ -4 & 4 & 0 & 7 \end{bmatrix}$

34. $\begin{bmatrix} 0 & 13 & 8 & 4 \\ 4 & 9 & 8 & 4 \\ 8 & 6 & 12 & 8 \\ 0 & 5 & 0 & -4 \end{bmatrix}$

35. $\begin{bmatrix} 11 & -6 & 4 & -10 & -4 \\ -3 & 5 & -2 & 4 & 1 \\ -8 & 12 & -3 & 12 & 4 \\ 1 & 6 & -2 & 3 & -1 \\ 8 & -18 & 8 & -14 & -1 \end{bmatrix}$