

Section 5.2: The Characteristic Equation

Chapter 5 : Eigenvalues and Eigenvectors Math 1554 Linear Algebra

Topics and	Objective
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TopicsWe will cover these topics in this section.

- $1. \ \, \hbox{The characteristic polynomial of a matrix}$
- 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.
- Characterize the long-term behaviour of dynamical systems using eigenvalue decompositions.

9/18 - 9/22 9/25 - 9/29

10/2 - 10/6

10/9 - 10/13

49

2.3,2.4

WS2.8,2.9 WS3.3,4.9

WS2.2,2.3

3.1,3.2 5.1.5.2

2.5

WS3.1,3.2 WS5.1,5.2 2.8

3.3

5.2

5.3

Exam 2, Review

Cancelled

WS2.4.2.5

The Characteristic Polynomial Recall:

Section 52 Stide 218

 λ is an eigenvalue of $A \Leftrightarrow (A - \lambda I)$ is not Therefore, to calculate the eigenvalues of A, we can solve

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

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 ______ of A.

Example

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

So the eigenvalues of \boldsymbol{A} are:

Characteristic Polynomial of 2×2 Matrices

Express the characteristic equation of

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

in terms of its determinant. What is the equation when ${\cal M}$ is singular?

Algebraic Multiplicity

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

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

Section 5.2 Slide 223

Section 5.2 Slide 222

Geometric Multiplicity

Definition

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

- 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.

Example

Give an example of a 4×4 matrix with $\lambda=0$ the only eigenvalue, but the geometric multiplicity of $\lambda=0$ is one.

Recall: Long-Term Behavior of Markov Chains

We often want to know what happens to a Markov Chain

$$\vec{x}_{k+1} = P\vec{x}_k, \quad k = 0, 1, 2, ...$$

as $k \to \infty$. $\,\circ\,$ If P is regular, then there is a $_$

- Now lets ask: \circ 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:

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



Goal: use eigenvalues to describe the long-term behavior of our system.

What are the corresponding eigenvectors of $P\mbox{\it ?}$

What are the eigenvalues of P?

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

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,

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

Section 5.2 Slide 230

Additional Examples (if time permits)

- 1. True or false.
 - a) If A is similar to the identity matrix, then A is equal to the identity matrix.
 - 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?

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

Section 5.2 Slide 231

5.2 Exercises

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

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

" [-4 4] " [2 3]

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 x 3 matrix is not easy to do with just row operations, because the variable A is involved.]

9.
$$\begin{bmatrix} 1 & 0 & -1 \\ 2 & 3 & -1 \\ 0 & 6 & 0 \end{bmatrix}$$
 10.
$$\begin{bmatrix} 0 & 3 & 1 \\ 3 & 0 & 2 \\ 1 & 2 & 0 \end{bmatrix}$$

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

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 λ below such that the eigenspace for λ = 5 is two-dimensional:

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

Let A be an n × n matrix, and suppose A has n real eigenvalues, λ₁,..., λ_n, repeated according to multiplicities, so that det(A − λI) = (λ₁ − λ)(λ₂ − λ)···(λ_n − λ)

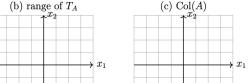
with an eigenvalue λ .

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

- (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
- (T/F) The matrix A and its transpose, A^T, have different sets of eigenvalues.
- 24. (T/F) The matrices A and B⁻¹AB 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.
- (T/F) If two matrices have the same set of eigenvalues, then they are similar.
- (T/F) If λ + 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 × n identity matrix is 1 with algebraic multiplicity n.
- (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 $x \to Ax$, $A \in \mathbb{R}^{2\times 2}$, 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? (a) Null(A)(b) range of T_A







Course Schedule

8/21 - 8/25 1.1 WS1.1 1.2

8/28 - 9/1

9/25 - 9/29

14 11/20 - 11/24 7.3.7.4 15 11/27 - 12/1 7.4

WS6.1 WS6.5.6.6 WS7.2.7.3 WS7.3.7.4

WS1.3,1.4 1.5

WS1.9,2.1 WS2.8.2.9

WS1.7

W\$3.1.3.2

1.3

- true false a) $S = {\vec{x} \in \mathbb{R}^3 | x_1 = a, x_2 = 4a, x_3 = x_1x_2}$ is a subspace for any $a \in \mathbb{R}$. 0 0 b) If A is square and non-zero, and $A\vec{x} = A\vec{y}$ for some $\vec{x} \neq \vec{y}$, then $\det(A) \neq 0$.
- 2. Indicate true if the statement is true, otherwise, indicate false.

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

(b)
$$A$$
 is 2×2 , $ColA$ is spanned by the vector $\begin{pmatrix} 2 \\ 3 \end{pmatrix}$ and $dim(Null(A)) = 0$. $A = \begin{pmatrix} \\ \\ \\ \end{pmatrix}$ (c) A is in RREF and $T_A : \mathbb{R}^3 \to \mathbb{R}^3$. The vectors u and v are a basis for the range of T .
$$u = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}, v = \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}, A = \begin{pmatrix} \\ \\ \\ \end{pmatrix}$$

P	F
possible	impossible

is also an eigenvector of
$$A$$
.

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

- 5. (2 points) Fill in the blanks.
- (a) If A is a 6×4 matrix in RREF and rank(A) = 4, what is the rank of A^T ?
 - (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 $\det(A)$?

- 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.

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Section 5.3: Diagonalization

Chapter 5: Eigenvalues and Eigenvectors

Math 1554 Linear Algebra

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

Topics and Objectives

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.

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

n 5.3 Slide 232 Section 5.3 Slide 233

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Section 5.3	Side 232									Secti	on 5.3 Slide 237	1									/30 - 11/3	6.4	W56.3,6.4	6.4,6.5	WS6		6.5	
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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

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

Diagonalization

Theorem

 $\overrightarrow{\text{If }A\text{ is diagonalizable}}\Leftrightarrow A\text{ has }n\text{ linearly independent eigenvectors}.$

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

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

$$A = [\vec{v}_1 \ \vec{v}_2 \cdots \vec{v}_n]$$

$$\begin{bmatrix} \lambda_1 \\ \lambda_2 \\ \end{bmatrix} [\vec{v}_1 \ \vec{v}_2 \cdots \vec{v}_n]^{-1}$$

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

Section 5.3 Single 23

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 × n
- $\bullet \ \ A \ \ \mbox{has distinct eigenvalues} \ \ \lambda_1, \ldots, \lambda_k, \ k \leq n$
- ullet $a_i = \mathsf{algebraic}$ multiplicity of λ_i
- $d_i = \text{dimension of } \lambda_i \text{ eigenspace ("geometric multiplicity")}$

Then

- hen
- $1. \ d_i \leq a_i \ \text{for all} \ i$
- 2. A is diagonalizable $\Leftrightarrow \Sigma d_i = n \Leftrightarrow d_i = a_i$ for all i
- 3. A is diagonalizable \Leftrightarrow the eigenvectors, for all eigenvalues, together form a basis for $\mathbb{R}^n.$

Section 5.3 Slide 240

Section 5.3 Slide 24

Diagonalize if possible.	Diagonalize if possible.
$\begin{pmatrix} 2 & 6 \\ 0 & -1 \end{pmatrix}$	$\begin{pmatrix} 3 & 1 \\ 0 & 3 \end{pmatrix}$

Example 2

ction 5.3 Slide 238 Section 5.3 Slide 239

Example 1

Example 3

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

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

Additional Example (if time permits)

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

generates a well-known sequence of numbers.

number in this sequence.

Use a diagonalization to find a matrix equation that gives the n^{th}

ection 5.3 Slide 242

Additional Example (if time permits)

generates a well-known sequence of numbers.

Note that

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

Use a diagonalization to find a matrix equation that gives the \boldsymbol{n}^{th}

number in this sequence.

Section 5.3 Slide 243

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

- Let A be an $n \times n$ matrix whose distinct eigenvalues are $\lambda_1, \ldots, \lambda_p$. a. For $1 \le k \le p$, the dimension of the eigenspace for λ_k is less than or equal to the multiplicity of the eigenvalue λ_k . b. The matrix \boldsymbol{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 .

- c. If A is diagonalizable and \mathcal{B}_k is a basis for the eigenspace corresponding to λ_k for each k, then the total collection of vectors in the sets $\mathcal{B}_1, \dots, \mathcal{B}_p$ 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.
$$\begin{bmatrix} a & 0 \\ 3(a-b) & b \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 3 & 1 \end{bmatrix} \begin{bmatrix} a & 0 \\ 0 & b \end{bmatrix} \begin{bmatrix} 1 \\ -3 \end{bmatrix}$$

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

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

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

$$\begin{bmatrix} -1 \\ -1 \\ 2 \end{bmatrix} \qquad \mathbf{14.} \begin{bmatrix} 4 & 0 & -2 \\ 2 & 5 & 4 \\ 0 & 0 & 5 \end{bmatrix}$$

15.
$$\begin{bmatrix} 7 & 4 & 16 \\ 2 & 5 & 8 \\ -2 & -2 & -5 \end{bmatrix}$$
 16.
$$\begin{bmatrix} 0 & -4 & -6 \\ -1 & 0 & -3 \\ 1 & 2 & 5 \end{bmatrix}$$

7.
$$\begin{bmatrix} 1 & 4 & 0 \\ 0 & 0 & 5 \end{bmatrix}$$
18.
$$\begin{bmatrix} 6 & 13 & -2 \\ 12 & 16 & 1 \end{bmatrix}$$
9.
$$\begin{bmatrix} 5 & -3 & 0 & 9 \\ 0 & 3 & 1 & -2 \\ 0 & 0 & 2 & 0 \\ 0 & 0 & 0 & 2 \end{bmatrix}$$
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*.

and some invertible matrix P.
b. If Rⁿ 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 x 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 twodimensional. 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 threedimensional. Is it possible that A is not diagonalizable? Justify your answer.
 - 27. Show that if A is both diagonalizable and invertible, then so
 - **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 $D_1 =$
- 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 $A = P_2 D P_2^{-1}$. 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.
- 0 6 4 33. 6 12 -21 0 0 7
- 4 -1011 -6 -4^{-} -35 -24 1
- -312 -812 4 -26 3 -1