

SECTION VIII B:

DIAGONALIZING

We have discussed at the beginning of this chapter how diagonal matrices are unusual in having simple closed forms for their powers; e.g.

$$D \equiv \begin{bmatrix} 2 & 0 \\ 0 & 3 \end{bmatrix} \rightarrow D^k = \begin{bmatrix} 2^k & 0 \\ 0 & 3^k \end{bmatrix},$$

$$k = 0, 1, 2, \dots$$

Almost as good as
to have

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

for some invertible P ; then

$$\begin{aligned} A^2 &= (PDP^{-1})(PDP^{-1}) = (PD)(P^{-1}P)(DP^{-1}) \\ &= (PD)(I)DP^{-1} = PD^2P^{-1} \end{aligned}$$

$$\begin{aligned} A^3 &= (PDP^{-1})(PD^2P^{-1}) = PD(P^{-1}P)D^2P^{-1} \\ &= PD^3P^{-1} \end{aligned}$$

⋮

TERMINOLOGY 8.18 ^{p. 687}

$$\begin{bmatrix} \lambda_1 & & 0 \\ & \lambda_2 & \\ 0 & \dots & \lambda_m \end{bmatrix} \text{ will mean}$$

an $(m \times m)$ diagonal matrix
whose diagonal entries, going
down the diagonal, are
 $\lambda_1, \lambda_2, \dots, \lambda_m$.

THEOREM 8.19

$$\text{If } A = P \begin{bmatrix} \lambda_1 & & 0 \\ & \lambda_2 & \\ 0 & \dots & \lambda_m \end{bmatrix} P^{-1},$$

then

$$A^k = P \begin{bmatrix} \lambda_1^k & & & \\ & \lambda_2^k & & \\ & & \circ & \\ & & \circ & \ddots \\ & & & \circ & \lambda_m^k \end{bmatrix} P^{-1}$$

for $k = 1, 2, 3, \dots$

As a corollary, we have the second of our two methods for solving Difference Equations.

This method works for any initial data \vec{x}_0 , for A as in Theorem 8.19, compare to 8.15.

GLOBAL METHOD for solving Difference Equations 8.20

If A is as in Theorem 8.19,
then, for any \vec{x}_0 ,

$$\vec{x}_n \equiv P \begin{bmatrix} \lambda_1^n & & & 0 \\ & \lambda_2^n & & \\ & & \ddots & \\ 0 & & & \lambda_m^n \end{bmatrix} P^{-1} \vec{x}_0 \quad \left(n = \begin{matrix} 1, 2, 3, \dots \end{matrix} \right)$$

is the solution of

$$\vec{x}_{k+1} = A \vec{x}_k \quad (k = 0, 1, 2, \dots)$$

Example 8.21

$$\text{If } A \equiv \begin{bmatrix} 4 & 2 \\ -1 & 1 \end{bmatrix}, \text{ then}$$

we will soon know how to show that

$$A = P \begin{bmatrix} 2 & 0 \\ 0 & 3 \end{bmatrix} P^{-1},$$

$$\text{with } P = \begin{bmatrix} 1 & 2 \\ -1 & -1 \end{bmatrix}, \quad P^{-1} = \begin{bmatrix} -1 & -2 \\ 1 & 1 \end{bmatrix}$$

(Construction of P & $\begin{bmatrix} 2 & 0 \\ 0 & 3 \end{bmatrix}$ soon to appear)

By Theorem 8.19, for p. 691
 $k = 1, 2, 3, \dots$,

$$A^k = P \begin{bmatrix} 2^k & 0 \\ 0 & 3^k \end{bmatrix} P^{-1} =$$

$$\begin{bmatrix} 1 & 2 \\ -1 & -1 \end{bmatrix} \begin{bmatrix} -(2^k) & -(2^{k+1}) \\ 3^k & 3^k \end{bmatrix} =$$

$$\begin{bmatrix} (-2^k) + 2(3^k) & (-(2^{k+1}) + 2(3^k)) \\ (2^k - (3^k)) & (2^{k+1} - (3^k)) \end{bmatrix}$$

By 8.20, the solution of

$$\vec{x}_{k+1} = \begin{bmatrix} 4 & 2 \\ -1 & 1 \end{bmatrix} \vec{x}_k \quad (k = 0, 1, 2, \dots)$$

$$\vec{x}_0 = \begin{bmatrix} 3 \\ -1 \end{bmatrix} \quad \text{is}$$

$$\vec{x}_n = A^n \vec{x}_0 = P \begin{bmatrix} 2^n & 0 \\ 0 & 3^n \end{bmatrix} P^{-1} \begin{bmatrix} 3 \\ -1 \end{bmatrix} =$$

$$\begin{bmatrix} (-(2^n) + 2(3^n)) & (-(2^{n+1}) + 2(3^n)) \\ (2^n - (3^n)) & (2^{n+1} - (3^n)) \end{bmatrix} \begin{bmatrix} 3 \\ -1 \end{bmatrix} =$$

$$\begin{bmatrix} 3(-(2^n) + 2(3^n)) - (-(2^{n+1}) + 2(3^n)) \\ 3(2^n - (3^n)) - (2^{n+1} - (3^n)) \end{bmatrix} =$$

$$\begin{bmatrix} -3(2^n) + 2^{n+1} - 2(3^n) + 2(3^{n+1}) \\ 3(2^n) - 2^{n+1} + 3^n - (3^{n+1}) \end{bmatrix}$$

$$(n=1, 2, 3, \dots)$$

We need a name for the hypothesis of Theorem 8.19. First, a more general name.

DEFINITION 8.22

Matrices A and B are

similar if

$$A = PBP^{-1},$$

for some invertible matrix P .

DEFINITION 8.23

A matrix A is

diagonalizable

if it is similar to a

diagonal matrix; that is,

$$(*) \quad A = P D P^{-1}$$

for some diagonal matrix D and invertible matrix P .

(*) is then a

diagonalization

of A .

Example 8.24

$$A \equiv \begin{bmatrix} 4 & 2 \\ -1 & 1 \end{bmatrix}, \text{ from}$$

Example 8.21, is diagonalizable, with

diagonalization

$$A = P D P^{-1},$$

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

PLEASE NOTE that

$$A \begin{bmatrix} 1 \\ -1 \end{bmatrix} = 2 \begin{bmatrix} 1 \\ -1 \end{bmatrix} \quad \& \quad A \begin{bmatrix} 2 \\ -1 \end{bmatrix} = 3 \begin{bmatrix} 2 \\ -1 \end{bmatrix};$$

P has eigenvectors for columns,
and D has corresponding
eigenvalue for diagonal entries.

HOW TO DIAGONALIZE 8.25

IF A is $(m \times m)$ and
 $\{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_m\}$ is a basis
 for \mathbb{R}^m , with

$$A \vec{v}_k = \lambda_k \vec{v}_k \quad (k=1, 2, \dots, m),$$

let $P \equiv [\vec{v}_1 \ \vec{v}_2 \ \dots \ \vec{v}_m]$,

$$D \equiv \begin{bmatrix} \lambda_1 \vec{e}_1 & & & \\ & \lambda_2 \vec{e}_2 & & \\ & & \dots & \\ & & & \lambda_m \vec{e}_m \end{bmatrix}$$

$$= \begin{bmatrix} \lambda_1 & & & \\ & \lambda_2 & & \\ & & \dots & \\ & & & \lambda_m \end{bmatrix}$$

Then P is invertible and

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

BEST WAY to get

$$\{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_m\} :$$

Take the union of the bases
for the eigenspaces;

the resulting set will
automatically be linearly
independent.

Partial Proof!

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For $1 \leq j \leq m$,

$$A\vec{v}_j = \lambda_j \vec{v}_j, \quad P\vec{e}_j = \vec{v}_j, \quad \text{thus}$$

$$P^{-1}\vec{v}_j = \vec{e}_j, \quad \text{so that}$$

$$(P^{-1}AP)\vec{e}_j = P^{-1}A\vec{v}_j =$$

$$P^{-1}(\lambda_j \vec{v}_j) = \lambda_j(P^{-1}\vec{v}_j) = \lambda_j \vec{e}_j$$

$$= D\vec{e}_j.$$

This implies that

$$P^{-1}AP = D, \quad \text{or}$$

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

Examples 8.26

In each of the following, use the information given to diagonalize A .

$$(1) A \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} = \begin{bmatrix} -5 \\ 0 \\ -5 \end{bmatrix}, \quad A \begin{bmatrix} 1 \\ 1 \\ 2 \end{bmatrix} = \begin{bmatrix} -5 \\ -5 \\ -10 \end{bmatrix}$$

$$A \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 7 \end{bmatrix}.$$

SOLUTION: $\vec{v}_1 \equiv (1, 0, 1)$, $\vec{v}_2 \equiv (1, 1, 2)$,

$\vec{v}_3 \equiv (0, 0, 1) \rightarrow A\vec{v}_1 = (-5)\vec{v}_1$, $A\vec{v}_2 = (-5)\vec{v}_2$

$A\vec{v}_3 = 7\vec{v}_3$, so define

$$P \equiv \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 0 \\ 1 & 2 & 1 \end{bmatrix}, \quad D \equiv \begin{bmatrix} (-5) & 0 & 0 \\ 0 & (-5) & 0 \\ 0 & 0 & 7 \end{bmatrix}$$

then $A = P D P^{-1}$.

OR we could've chosen

$$P \equiv \begin{bmatrix} 1 & 1 & 0 \\ 1 & 0 & 0 \\ 2 & 1 & 1 \end{bmatrix}, \text{ same } D$$

$$\text{OR } P \equiv \begin{bmatrix} 1 & 0 & 1 \\ 0 & 0 & 1 \\ 1 & 1 & 2 \end{bmatrix}, D \equiv \begin{bmatrix} (-5) & 0 & 0 \\ 0 & 7 & 0 \\ 0 & 0 & (-5) \end{bmatrix}$$

More than one answer is possible, but in each column, the eigenvalue in D must correspond to the eigenvector in P .

We should've first checked $\text{rank} [\vec{v}_1 \ \vec{v}_2 \ \vec{v}_3] = 3$, to

ensure $\{\vec{v}_1, \vec{v}_2, \vec{v}_3\}$ is a basis for \mathbb{R}^3 .

(2) A has eigenspaces

$$E_0 = \left\{ \begin{bmatrix} s+t \\ 2s \\ 0 \\ t \end{bmatrix} \mid s, t \text{ real} \right\}$$

$$E_1 = \left\{ \begin{bmatrix} t \\ 3t \\ 0 \\ 0 \end{bmatrix} \mid t \text{ real} \right\} \quad \text{and}$$

$$E_{(-1)} = \left\{ \begin{bmatrix} 0 \\ 0 \\ t \\ 0 \end{bmatrix} \mid t \text{ real} \right\}$$

SOLUTION: Get bases for the eigenspaces.

$$E_0: \begin{bmatrix} s+t \\ 2s \\ 0 \\ t \end{bmatrix} = s \begin{bmatrix} 1 \\ 2 \\ 0 \\ 0 \end{bmatrix} + t \begin{bmatrix} 1 \\ 0 \\ 0 \\ 1 \end{bmatrix}$$

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$$\text{basis} \left\{ \begin{bmatrix} 1 \\ 2 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ 0 \\ 1 \end{bmatrix} \right\}$$

$$E_1: \begin{bmatrix} t \\ 3t \\ 0 \\ 0 \end{bmatrix} = t \begin{bmatrix} 1 \\ 3 \\ 0 \\ 0 \end{bmatrix}$$

$$\text{basis} \left\{ \begin{bmatrix} 1 \\ 3 \\ 0 \\ 0 \end{bmatrix} \right\}$$

$$E_{(-1)}: \begin{bmatrix} 0 \\ 0 \\ t \\ 0 \end{bmatrix} = t \begin{bmatrix} 0 \\ 0 \\ 1 \\ 0 \end{bmatrix}$$

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

For P , paste those eigenspace bases

together:

$$P \equiv \begin{bmatrix} 1 & 1 & 1 & 0 \\ 2 & 0 & 3 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 1 & 0 & 0 \end{bmatrix}$$

Corresponding eigenvalues

for $D = \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & -1 \end{bmatrix}$,

then $A = PDP^{-1}$

(3) $A \equiv \begin{bmatrix} 0 & 4 \\ -1 & 5 \end{bmatrix}$.

SOLUTION: Since we don't know eigenvectors, we must begin by finding eigenvalues.

STEP ONE: $0 = C_A(t) \equiv \det(A - tI)$

$$= \det \begin{bmatrix} -t & 4 \\ -1 & (5-t) \end{bmatrix} = t(t-5) - (-4)$$

$$= t^2 - 5t + 4 = (t-1)(t-4)$$

→ eigenvalues 1, 4.

STEP TWO: eigenspaces

$$E_1: [(A-I) | \vec{0}] = \left[\begin{array}{cc|c} -1 & 4 & 0 \\ -1 & 4 & 0 \end{array} \right] \begin{array}{l} R_2 - R_1 \\ \rightarrow \end{array}$$

$$\left[\begin{array}{cc|c} -1 & 4 & 0 \\ 0 & 0 & 0 \end{array} \right] \rightarrow -x_1 + 4x_2 = 0 \rightarrow$$

$$E_1 = \left\{ \begin{bmatrix} 4x_2 \\ x_2 \end{bmatrix} \mid x_2 \text{ real} \right\}$$

$$E_4: [(A-4I) | \vec{0}] = \left[\begin{array}{cc|c} -4 & 4 & 0 \\ -1 & 1 & 0 \end{array} \right] \begin{array}{l} R_1 \leftrightarrow R_2 \\ \rightarrow \end{array}$$

$$\left[\begin{array}{cc|c} -1 & 1 & 0 \\ -4 & 4 & 0 \end{array} \right] \begin{array}{l} R_2 - 4R_1 \\ \rightarrow \end{array} \left[\begin{array}{cc|c} -1 & 1 & 0 \\ 0 & 0 & 0 \end{array} \right] \rightarrow$$

$$-x_1 + x_2 = 0 \rightarrow E_4 = \left\{ \begin{bmatrix} x_2 \\ x_2 \end{bmatrix} \mid x_2 \text{ real} \right\}$$

STEP THREE: bases

$$E_1: \begin{bmatrix} 4x_2 \\ x_2 \end{bmatrix} = x_2 \begin{bmatrix} 4 \\ 1 \end{bmatrix}; \text{ basis } \left\{ \begin{bmatrix} 4 \\ 1 \end{bmatrix} \right\}$$

$$E_4: \begin{bmatrix} x_2 \\ x_2 \end{bmatrix} = x_2 \begin{bmatrix} 1 \\ 1 \end{bmatrix}; \text{ basis } \left\{ \begin{bmatrix} 1 \\ 1 \end{bmatrix} \right\}$$

STEP FOUR: paste

$$P = \begin{bmatrix} 4 & 1 \\ 1 & 1 \end{bmatrix}, \quad D = \begin{bmatrix} 1 & 0 \\ 0 & 4 \end{bmatrix}$$

OR

$$P = \begin{bmatrix} 1 & 4 \\ 1 & 1 \end{bmatrix}, \quad D = \begin{bmatrix} 4 & 0 \\ 0 & 1 \end{bmatrix}$$

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

Example 8.27

$$\text{Let } A = \begin{bmatrix} 1 & -1 & -1 \\ -1 & 1 & -1 \\ 1 & 1 & 3 \end{bmatrix},$$

from Example 8.16(c), page 673.

Solve each of the following.

$$(1) \vec{x}_{k+1} = A \vec{x}_k, \quad \vec{x}_0 = (1, 0, 0).$$

$$(2) \vec{x}_{k+1} = A \vec{x}_k, \quad \vec{x}_0 = (0, 1, 0)$$

$$(3) \vec{x}_{k+1} = A \vec{x}_k, \quad \vec{x}_0 = (0, 0, 1)$$

$$(4) \vec{x}_{k+1} = A \vec{x}_k, \quad \vec{x}_0 = (0, 1, -1).$$

SOLUTIONS:

For the global method
8.20 or the local method
8.15 (we used this in Example
8.16), we need eigenvectors
and eigenvalues. In Example
8.16(c), we were given
eigenvectors

$$(*) \quad \vec{v}_1 \equiv \begin{bmatrix} -1 \\ -1 \\ 1 \end{bmatrix}, \quad \vec{v}_2 \equiv \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix}, \quad \vec{v}_3 \equiv \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}$$

and used those to get
eigenvalues:

$$A\vec{v}_1 = \vec{v}_1, \quad A\vec{v}_2 = 2\vec{v}_2,$$

$$(\text{**}) \quad A\vec{v}_3 = 2\vec{v}_3.$$

For this problem (compare to Example 8.16(c)), we prefer 8.20. 8.15 requires writing \vec{x}_0 as a linear combination of $\{\vec{v}_1, \vec{v}_2, \vec{v}_3\}$; we would have to do this separately for each of the \vec{x}_0 s in (1), (2), (3), and (4).

For the global method 8.20 we calculate A^n , then

apply it to each \vec{x}_0

leaving only matrix multiplication.

Getting A^n begins with diagonalizing A :

From (*), $P \equiv \begin{bmatrix} -1 & -1 & -1 \\ -1 & 1 & 0 \\ 1 & 0 & 1 \end{bmatrix}$;

From (**),

$$D \equiv \begin{bmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 2 \end{bmatrix} \quad (\text{see 8.25}),$$

then $A = PDP^{-1}$ (diagonalized)

and, by (8.19),

$$(***) \quad A^n = P \begin{bmatrix} 1^n & 0 & 0 \\ 0 & 2^n & 0 \\ 0 & 0 & 2^n \end{bmatrix} P^{-1} \quad (n = 0, 1, 2, \dots)$$

For P^{-1} : $[P | I] =$

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$$\left[\begin{array}{ccc|ccc} -1 & -1 & -1 & 1 & 0 & 0 \\ -1 & 1 & 0 & 0 & 1 & 0 \\ 1 & 0 & 1 & 0 & 0 & 1 \end{array} \right] \rightarrow (R_1 \leftrightarrow R_3)$$

$$\left[\begin{array}{ccc|ccc} 1 & 0 & 1 & 0 & 0 & 1 \\ -1 & 1 & 0 & 0 & 1 & 0 \\ -1 & -1 & -1 & 0 & 0 & 0 \end{array} \right] \rightarrow \begin{pmatrix} R_2 + R_1 \\ R_3 + R_1 \end{pmatrix}$$

$$\left[\begin{array}{ccc|ccc} 1 & 0 & 1 & 0 & 0 & 1 \\ 0 & 1 & 1 & 0 & 1 & 1 \\ 0 & -1 & 0 & 1 & 0 & 1 \end{array} \right] \rightarrow (R_3 + R_2)$$

$$\left[\begin{array}{ccc|ccc} 1 & 0 & 1 & 0 & 0 & 1 \\ 0 & 1 & 1 & 0 & 1 & 1 \\ 0 & 0 & 1 & 1 & 1 & 2 \end{array} \right] \rightarrow \begin{pmatrix} R_1 - R_3 \\ R_2 - R_3 \end{pmatrix}$$

$$\left[\begin{array}{ccc|ccc} 1 & 0 & 0 & -1 & -1 & -1 \\ 0 & 1 & 0 & -1 & 0 & -1 \\ 0 & 0 & 1 & -1 & 1 & 2 \end{array} \right] \rightarrow P^{-1} = \begin{bmatrix} -1 & -1 & -1 \\ -1 & 0 & -1 \\ 1 & 1 & 2 \end{bmatrix}.$$

Now we must multiply a p. 711
 great deal in (~~***~~)

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

$$\begin{bmatrix} -1 & -1 & -1 \\ -1 & 1 & 0 \\ 1 & 0 & 1 \end{bmatrix} \begin{bmatrix} -1 & -1 & -1 \\ -(2^n) & 0 & -(2^n) \\ 2^n & 2^n & 2^{n+1} \end{bmatrix} =$$

$$\begin{bmatrix} (1 + \cancel{2^n} - \cancel{(2^n)}) & (1 - (2^n)) & (1 + 2^n - (2^{n+1})) \\ (1 - (2^n)) & 1 & (1 - (2^n)) \\ (-1 + 2^n) & (-1 + 2^n) & (-1 + 2^{n+1}) \end{bmatrix}$$

$$(1) \vec{x}_n = A^n \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} = \begin{bmatrix} 1 \\ (1 - (2^n)) \\ (-1 + 2^n) \end{bmatrix}$$

$$(2) \vec{x}_n = A^n \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} = \begin{bmatrix} (1 - (2^n)) \\ 1 \\ (-1 + 2^n) \end{bmatrix}$$

$$(3) \vec{x}_n = A^n \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} = \begin{bmatrix} (1 + 2^n - (2^{n+1})) \\ (1 - (2^n)) \\ (-1 + 2^{n+1}) \end{bmatrix}$$

$$(4) \vec{x}_n = A^n \begin{bmatrix} 0 \\ 1 \\ -1 \end{bmatrix} = \begin{matrix} \text{(answer to (2))} \\ - \text{(answer to (3))} \end{matrix}$$

$$= \begin{bmatrix} (1 - 2^n) - (1 + 2^n - (2^{n+1})) \\ 1 - (1 - (2^n)) \\ (-1 + 2^n) - (-1 + 2^{n+1}) \end{bmatrix}$$

$$= \begin{bmatrix} 0 \\ 2^n \\ 2^n - (2^{n+1}) \end{bmatrix}$$

Now we have some bad news: not every square matrix is diagonalizable.

This will be easier to demonstrate after a general statement about similar matrices.

PROPOSITION 8.28

If A and B are similar, then their characteristic polynomials C_A and C_B are equal; in particular, A and B have the same eigenvalues.

Proof: If $A = PBP^{-1}$,

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then

$$C_A(t) \equiv \det(A - tI) =$$

$$\det(PBP^{-1} - P t I P^{-1}) =$$

$$\det(P(B - tI)P^{-1}) =$$

$$(\det P) (\det(B - tI)) (\det(P^{-1}))$$

$$= \det(B - tI) \equiv C_B(t),$$

$$\text{since } (\det P)(\det(P^{-1})) = \det(PP^{-1})$$

$$= \det(I) = 1.$$

Equality of $\{\text{eigenvalues}\}$ now follows from Theorem 8.8,

Example 8.29

We claim that $A \equiv \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}$
is not diagonalizable.

We will prove this claim
"by contradiction," meaning
we pretend the claim is false,
then arrive at a contradiction as
a consequence of our pretense.

Suppose, "for the sake of contra-
diction," that A is diagonalizable.
Then $A = PDP^{-1}$, for some
invertible P & diagonal D ;

$$D = \begin{bmatrix} a & 0 \\ 0 & b \end{bmatrix}, \text{ for}$$

some numbers a, b .

$$C_A(t) = \det \begin{bmatrix} -t & 1 \\ 0 & -t \end{bmatrix} = t^2;$$

by Proposition 8.28,

$$t^2 = C_D(t) = \det \begin{bmatrix} (a-t) & 0 \\ 0 & (b-t) \end{bmatrix}$$

$$= (a-t)(b-t), \text{ thus } a = 0 = b,$$

so that

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

$$\begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}, \text{ a contradiction}$$

of the fact that

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

Thus our pretense is false:

A is not diagonalizable.

In diagonalizing examples we've seen eigenvalues appear more than once in the diagonal matrix similar to the matrix of interest.

There are two natural ways to define this repetition, or multiplicity,

informally, "how many times" an eigenvalue appears.

DEFINITIONS 8.30

Suppose λ is an eigenvalue of the square matrix A .

The **geometric multiplicity** of λ is $\dim(E_\lambda)$, the dimension of the eigenspace for λ .

The algebraic

multiplicity of λ

is the number of times
the factor $(\lambda - t)$ appears
in the factorization of
the characteristic polynomial
 $C_A(t)$.

Example 8.31

Let $A \equiv \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$, as in

Example 8.29.

We have already
calculated

$$C_A(t) = t^2,$$

thus, the algebraic multiplicity
of the eigenvalue 0 is 2.

For the geometric multiplicity
we need the eigenspace

$E_0 = \mathcal{N}(A)$: solving

$$[A | \vec{0}] = \left[\begin{array}{cc|c} 0 & 1 & 0 \\ 0 & 0 & 0 \end{array} \right] \rightarrow x_2 = 0;$$

$$E_0 = \left\{ \begin{bmatrix} x_1 \\ 0 \end{bmatrix} \mid x_1 \text{ is real} \right\},$$

so that

the geometric
multiplicity of the eigenvalue
0 is $\dim(E_0) = 1$.

The difference in
multiplicities of an eigenvalue
for A is equivalent to
 A being not diagonalizable.

THEOREM 8.32

Suppose A is an $(m \times m)$
matrix. Then the following
are equivalent.

(a) A is diagonalizable. ^{p. 722}

(b) For every eigenvalue λ , the geometric multiplicity of λ equals the algebraic multiplicity of λ .

(c) The sum of the dimensions of the eigenspaces equals n .

(d) There is a basis for \mathbb{R}^n consisting entirely of eigenvectors for A .

The intuition is that we need plenty of eigenvectors, enough to construct invertible P in the definition of diagonalizable (Definition 8.23)

Example 8.33

In each of the following, diagonalize A if possible.

$$(1) \quad C_A(t) = (t+1)(t-3)^2,$$

$$E_{(-1)} = \left\{ \begin{bmatrix} t \\ 0 \\ -t \end{bmatrix} \mid t \text{ real} \right\},$$

$$E_3 = \left\{ \begin{bmatrix} 0 \\ t \\ 0 \end{bmatrix} \mid t \text{ real} \right\}$$

SOLUTION:

algebraic multiplicity of
eigenvalue 3 equals

$2 \neq 1 =$ geometric multiplicity
of eigenvalue 3, so not

diagonalizable

by (b) of Theorem 8.32.

OR we could've added

$$\dim(E_{-1}) + \dim(E_3) = 1 + 1$$

$$= 2 \neq 3, \text{ \& } A \text{ is } (3 \times 3),$$

so we could invoke (c)

of Theorem 8.32.

(2) SAME as (1), except

$$E_3 = \left\{ \begin{bmatrix} s \\ t \\ 2s \end{bmatrix} \mid s, t \text{ real} \right\}$$

SOLUTION: basis for

$$E_{(-1)} : \left\{ \begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix} \right\}; \text{ for}$$

$$E_3 : \left\{ \begin{bmatrix} 1 \\ 0 \\ 2 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} \right\}.$$

$$P \equiv \begin{bmatrix} 1 & 1 & 0 \\ 0 & 0 & 1 \\ -1 & 2 & 0 \end{bmatrix}, \quad D \equiv \begin{bmatrix} -1 & 0 & 0 \\ 0 & 3 & 0 \\ 0 & 0 & 3 \end{bmatrix}$$

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

(3) A is 4×4 and

has eigenspaces

$$E_2 = \left\{ (0, s+t, 2t+s, t) \mid \begin{array}{l} s, t \\ \text{real} \end{array} \right\}$$

$$E_5 = \left\{ (t, 0, 0, 0) \mid t \text{ real} \right\}$$

(only eigenvalues are 2 and 5).

SOLUTION: $\dim(E_2) + \dim(E_5)$

$$= 2 + 1 = 3 \neq 4, \text{ and } A \text{ is } 4 \times 4,$$

so

not diagonalizable

Here are two special cases where it is easy to determine a matrix is diagonalizable.

THEOREM 8.34

If A is an $(n \times n)$ matrix, then A is diagonalizable if either of the following holds:

(a) A is symmetric; or

(b) A has n distinct eigenvalues.

Examples 8.35

Each of the following matrices is guaranteed by Theorem 8.34 to be diagonalizable.

(1) A is 4×4 and has eigenvalues $-2, 0, \pi, \sqrt{19}$

$$(2) A = \begin{bmatrix} 7 & \sqrt{\pi} & 0 \\ \sqrt{\pi} & \sqrt{2} & 9 \\ 0 & 9 & 19 \end{bmatrix}.$$

REMARK 8.36

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It can be shown that an $(m \times m)$ matrix A is symmetric if and only if \mathbb{R}^m has an orthogonal basis consisting entirely of eigenvectors for A ; compare to (d) of Theorem 8.32.

Example 8.37

For our final example, we would like to solve the fox and rabbit Difference Equation in Examples 1.25(4) and 7.22(b).

More than this, we would like to generalize Example 1.25(4) by having arbitrary specified rates of reproduction and (for the foxes) consumption.

As in Example 1.25(4), for $k = 0, 1, 2, \dots$

$r_k \equiv$ number of rabbits k years from now

$f_k \equiv$ number of foxes k years from now.

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In the following recursive model for interacting (a term probably made up by the foxes; the rabbits might have a less tolerant description)

foxes and rabbits, the number F represents fox fertility, R rabbit fertility, and E (for "Eating") is the number of rabbits eaten by each fox in a year.

$$(*) \begin{cases} r_{k+1} = R r_k - E f_k \\ f_{k+1} = F f_k \end{cases} \quad \begin{pmatrix} k=0, 1, \\ 2, \dots \end{pmatrix}$$

If $A \equiv \begin{bmatrix} R & -E \\ 0 & F \end{bmatrix}$, then

$$\begin{bmatrix} r_{k+1} \\ f_{k+1} \end{bmatrix} = A \begin{bmatrix} r_k \\ f_k \end{bmatrix} \quad (k=0, 1, 2, \dots)$$

thus

$$(**) \begin{bmatrix} r_n \\ f_n \end{bmatrix} = A^n \begin{bmatrix} r_0 \\ f_0 \end{bmatrix} \quad \begin{pmatrix} n=0, 1, 2, \\ \dots \end{pmatrix}$$

By counting the number of foxes and rabbits now, we can state the number of foxes and rabbits at any time in the future.

Note that Example 1.25(4) is the special case
 $F = 4$, $R = 100$, $E = 360$.

CASE I: $F \neq R$.

We will leave it to the reader to calculate that

the eigenvalues for A
are F and R , with
eigenspaces

$$E_R = \text{span} \left\{ \begin{bmatrix} 1 \\ 0 \end{bmatrix} \right\},$$

$$E_F = \text{span} \left\{ \begin{bmatrix} 1 \\ \epsilon \end{bmatrix} \right\}, \text{ where}$$

$$\epsilon \equiv \left(\frac{R-F}{E} \right).$$

Use 8.25 to diagonalize A :

$$P \equiv \begin{bmatrix} 1 & 1 \\ 0 & \epsilon \end{bmatrix}, \quad D \equiv \begin{bmatrix} R & 0 \\ 0 & F \end{bmatrix}$$

$$\rightarrow A = P D P^{-1}.$$

We also leave it to
the reader to calculate

$$P^{-1} = \frac{1}{\epsilon} \begin{bmatrix} \epsilon & -1 \\ 0 & 1 \end{bmatrix}.$$

Now we may get powers
of A . For $n = 0, 1, 2, \dots$

$$A^n = P \begin{bmatrix} R^n & 0 \\ 0 & F^n \end{bmatrix} P^{-1} =$$

$$\frac{1}{\epsilon} \begin{bmatrix} 1 & 1 \\ 0 & \epsilon \end{bmatrix} \begin{bmatrix} \epsilon(R^n) & -(R^n) \\ 0 & F^n \end{bmatrix} =$$

$$\frac{1}{\epsilon} \begin{bmatrix} \epsilon(R^n) & -(R^n) + F^n \\ 0 & \epsilon(F^n) \end{bmatrix} =$$

$$\begin{bmatrix} R^n & \left(\frac{E}{R-F}\right)(F^n - (R^n)) \\ 0 & F^n \end{bmatrix} \cdot$$

By $(**)$,

$$\begin{bmatrix} r_n \\ f_n \end{bmatrix} = \begin{bmatrix} R^n r_0 + \left(\frac{E}{R-F}\right)(F^n - (R^n)) f_0 \\ F^n f_0 \end{bmatrix}$$

is the solution of $(*)$, at least until extinction.

WHAT ABOUT Extinction?

Extinction occurs by the n^{th} year from now

if and only if

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$$r_n = \left[R^n r_0 + \left(\frac{E}{R-F} \right) (F^n - (R^n)) f_0 \right]$$

≤ 0 if and only if (after some algebra)

$$\frac{r_0}{f_0} \leq \left(\frac{E}{R-F} \right) \left(1 - \left(\frac{F}{R} \right)^n \right).$$

It is interesting that extinction depends entirely on the initial ratio of rabbits to foxes.

NOTE that, if $F > R$ (foxes more fertile than rabbits), then extinction is inevitable, since then

$$\left| \left(\frac{E}{R-F} \right) \left(1 - \left(\frac{F}{R} \right)^n \right) \right|$$

get arbitrarily large, as n get large.

If $F < R$, then $\left(\frac{E}{R-F} \right) \left(1 - \left(\frac{F}{R} \right)^n \right)$ is increasing to

$$\left(\frac{E}{R-F} \right)$$
 as n increase, to ∞ ,

thus

Extinction occurs eventually
if and only if

$$\frac{r_0}{f_0} < \left(\frac{E}{R-F} \right).$$

So long as rabbits are
more fertile than foxes,
extinction can be avoided
by starting out with enough
rabbits (relative to the
number of foxes).

CASE II: $F = R$.

We will leave it to the reader to show that

$$A \equiv \begin{bmatrix} F & -E \\ 0 & F \end{bmatrix}$$

is not diagonalizable.

But a pattern for powers of A does emerge:

$$A^2 = \begin{bmatrix} F^2 & -2EF \\ 0 & F^2 \end{bmatrix}, \quad A^3 = \begin{bmatrix} F^3 & -3EF^2 \\ 0 & F^3 \end{bmatrix},$$

...

the pattern turns out to be

$$A^n = \begin{bmatrix} F^n & -nEF^{n-1} \\ 0 & F^n \end{bmatrix} =$$

$$F^{n-1} \begin{bmatrix} F & -nE \\ 0 & F \end{bmatrix}, \text{ so that, by (**),}$$

$$\begin{bmatrix} r_n \\ f_n \end{bmatrix} = F^{n-1} \begin{bmatrix} F & -nE \\ 0 & F \end{bmatrix} \begin{bmatrix} r_0 \\ f_0 \end{bmatrix}$$

$$= F^{n-1} \begin{bmatrix} Fr_0 - nEf_0 \\ Ff_0 \end{bmatrix}$$

This implies that $r_n \leq 0$
 (extinction by n^{th} year)
 if and only if

$$\frac{r_0}{f_0} \leq \frac{nE}{F},$$

which is guaranteed to happen eventually, for any r_0, F_0 .

In CASE II, eventual extinction is guaranteed. It is not sufficient to have equal fertility of rabbits and foxes, if one wishes to avoid extinction; we must have rabbit fertility (strictly) greater than fox fertility.

Specific Example

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B.38

(1) In Example 1.25(4)

(*) of Example B.37 with
 $F = 4, R = 100, E = 360$

$$\left(\frac{E}{R-F}\right) = 3.75, \text{ thw}$$

extinction occurs eventually
if and only if

$$\left(\frac{r_0}{f_0}\right) < 3.75 ;$$

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if there are initially
100 foxes, you need 375
or more rabbits initially to avoid
extinction.

$$(2) \quad r_{k+1} = 2r_k - f_k$$

$$f_{k+1} = 3f_k$$

is guaranteed for eventual

extinction, regardless of

r_0 and f_0 , since $F = 3 \geq 2 = R$

in (*) of Example B.37.

$$(3) \quad r_{k+1} = 6r_k - 2f_k$$

$$f_{k+1} = 4f_k$$

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(*) with $R = 6$, $E = 2$ & $F = 4$

will avoid extinction if and

only if $\frac{r_0}{f_0} \geq 1$;

that is, the number of rabbits
is not less than the number
of foxes, since

$$\left(\frac{E}{R-F} \right) = 1$$

$$(4) \quad r_{k+1} = 6r_k - 200f_k$$

$$f_{k+1} = 4f_k$$

$$\left(\begin{array}{l} (*) \text{ with } R = 6, E = 200 \\ \text{d } F = 4 \end{array} \right)$$

will avoid extinction of

and only if $\frac{r_0}{f_0} \geq 100,$

since $\left(\frac{E}{R-F} \right) = 100.$

(4) has the same fertilitoes as (3), but hungrier foxes, hence we need more rabbits initially.