

3. a. Let  $H$  stand for "Healthy" and  $I$  stand for "Ill." Then the students' conditions are given by the table

From:		To:
H	I	
.95	.45	H
.05	.55	I

so the stochastic matrix is  $P = \begin{bmatrix} .95 & .45 \\ .05 & .55 \end{bmatrix}$ .

- b. Since 20% of the students are ill on Monday, the initial state vector is  $\mathbf{x}_0 = \begin{bmatrix} .8 \\ .2 \end{bmatrix}$ . For Tuesday's percentages, we calculate  $\mathbf{x}_1$ ; for Wednesday's percentages, we calculate  $\mathbf{x}_2$ :

$$\mathbf{x}_1 = P\mathbf{x}_0 = \begin{bmatrix} .95 & .45 \\ .05 & .55 \end{bmatrix} \begin{bmatrix} .8 \\ .2 \end{bmatrix} = \begin{bmatrix} .85 \\ .15 \end{bmatrix}$$

$$\mathbf{x}_2 = P\mathbf{x}_1 = \begin{bmatrix} .95 & .45 \\ .05 & .55 \end{bmatrix} \begin{bmatrix} .85 \\ .15 \end{bmatrix} = \begin{bmatrix} .875 \\ .125 \end{bmatrix}$$

Thus 15% of the students are ill on Tuesday, and 12.5% are ill on Wednesday.

- c. Since the student is well today, the initial state vector is  $\mathbf{x}_0 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$ . We calculate  $\mathbf{x}_2$ :

$$\mathbf{x}_1 = P\mathbf{x}_0 = \begin{bmatrix} .95 & .45 \\ .05 & .55 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} .95 \\ .05 \end{bmatrix}$$

$$\mathbf{x}_2 = P\mathbf{x}_1 = \begin{bmatrix} .95 & .45 \\ .05 & .55 \end{bmatrix} \begin{bmatrix} .95 \\ .05 \end{bmatrix} = \begin{bmatrix} .925 \\ .075 \end{bmatrix}$$

Thus the probability that the student is well two days from now is .925.

4. a. Let  $G$  stand for good weather,  $I$  for indifferent weather, and  $B$  for bad weather. Then the change in the weather is given by the table

From:			To:
G	I	B	
.6	.4	.4	G
.3	.3	.5	I
.1	.3	.1	B

so the stochastic matrix is  $P = \begin{bmatrix} .6 & .4 & .4 \\ .3 & .3 & .5 \\ .1 & .3 & .1 \end{bmatrix}$ .

- b. The initial state vector is  $\begin{bmatrix} .5 \\ .5 \\ 0 \end{bmatrix}$ . We calculate  $\mathbf{x}_1$ :

$$\mathbf{x}_1 = P\mathbf{x}_0 = \begin{bmatrix} .6 & .4 & .4 \\ .3 & .3 & .5 \\ .1 & .3 & .1 \end{bmatrix} \begin{bmatrix} .5 \\ .5 \\ 0 \end{bmatrix} = \begin{bmatrix} .5 \\ .3 \\ .2 \end{bmatrix}$$

12. From Exercise 2,  $P = \begin{bmatrix} .5 & .25 & .25 \\ .25 & .5 & .25 \\ .25 & .25 & .5 \end{bmatrix}$ , so  $P - I = \begin{bmatrix} -.5 & .25 & .25 \\ .25 & -.5 & .25 \\ .25 & .25 & -.5 \end{bmatrix}$ . Solving  $(P - I)\mathbf{x} = \mathbf{0}$  by row

reducing the augmented matrix gives

$$\left[ \begin{array}{ccc|c} -.5 & .25 & .25 & 0 \\ .25 & -.5 & .25 & 0 \\ .25 & .25 & -.5 & 0 \end{array} \right] \sim \left[ \begin{array}{ccc|c} 1 & 0 & -1 & 0 \\ 0 & 1 & -1 & 0 \\ 0 & 0 & 0 & 0 \end{array} \right]$$

Thus  $\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = x_3 \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$ , and one solution is  $\begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$ . Since the entries in  $\begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$  sum to 3, multiply by  $1/3$

to obtain the steady-state vector  $\mathbf{q} = \begin{bmatrix} 1/3 \\ 1/3 \\ 1/3 \end{bmatrix} = \begin{bmatrix} .333 \\ .333 \\ .333 \end{bmatrix}$ . Thus in the long run each food will be preferred equally.

13. a. From Exercise 3,  $P = \begin{bmatrix} .95 & .45 \\ .05 & .55 \end{bmatrix}$ , so  $P - I = \begin{bmatrix} -.05 & .45 \\ .05 & -.45 \end{bmatrix}$ . Solving  $(P - I)\mathbf{x} = \mathbf{0}$  by row reducing the augmented matrix gives

$$\left[ \begin{array}{cc|c} -.05 & .45 & 0 \\ .05 & -.45 & 0 \end{array} \right] \sim \left[ \begin{array}{cc|c} 1 & -9 & 0 \\ 0 & 0 & 0 \end{array} \right]$$

Thus  $\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = x_2 \begin{bmatrix} 9 \\ 1 \end{bmatrix}$ , and one solution is  $\begin{bmatrix} 9 \\ 1 \end{bmatrix}$ . Since the entries in  $\begin{bmatrix} 9 \\ 1 \end{bmatrix}$  sum to 10, multiply by

to obtain the steady-state vector  $\mathbf{q} = \begin{bmatrix} 9/10 \\ 1/10 \end{bmatrix} = \begin{bmatrix} .9 \\ .1 \end{bmatrix}$ .

- b. After many days, a specific student is ill with probability .1, and it does not matter whether that student is ill today or not.

14. From Exercise 4,  $P = \begin{bmatrix} .6 & .4 & .4 \\ .3 & .3 & .5 \\ .1 & .3 & .1 \end{bmatrix}$ , so  $P - I = \begin{bmatrix} -.4 & .4 & .4 \\ .3 & -.7 & .5 \\ .1 & .3 & -.9 \end{bmatrix}$ . Solving  $(P - I)\mathbf{x} = \mathbf{0}$  by row reducing the augmented matrix gives

$$\left[ \begin{array}{ccc|c} -.4 & .4 & .4 & 0 \\ .3 & -.7 & .5 & 0 \\ .1 & .3 & -.9 & 0 \end{array} \right] \sim \left[ \begin{array}{ccc|c} 1 & 0 & -3 & 0 \\ 0 & 1 & -2 & 0 \\ 0 & 0 & 0 & 0 \end{array} \right]$$

Thus  $\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = x_3 \begin{bmatrix} 3 \\ 2 \\ 1 \end{bmatrix}$ , and one solution is  $\begin{bmatrix} 3 \\ 2 \\ 1 \end{bmatrix}$ . Since the entries in  $\begin{bmatrix} 3 \\ 2 \\ 1 \end{bmatrix}$  sum to 6, multiply by  $1/6$  to

obtain the steady-state vector  $\mathbf{q} = \begin{bmatrix} 1/2 \\ 1/3 \\ 1/6 \end{bmatrix} = \begin{bmatrix} .5 \\ .333 \\ .167 \end{bmatrix}$ . Thus in the long run the chance that a day has good weather is 50%.

## 5.3 SOLUTIONS

1.  $P = \begin{bmatrix} 5 & 7 \\ 2 & 3 \end{bmatrix}$ ,  $D = \begin{bmatrix} 2 & 0 \\ 0 & 1 \end{bmatrix}$ ,  $A = PDP^{-1}$ , and  $A^k = PD^kP^{-1}$ . We compute  $P^{-1} = \begin{bmatrix} 3 & -7 \\ -2 & 5 \end{bmatrix}$ ,  $D^k = \begin{bmatrix} 16 & 0 \\ 0 & 1 \end{bmatrix}$ .

and  $A^k = \begin{bmatrix} 5 & 7 \\ 2 & 3 \end{bmatrix} \begin{bmatrix} 16 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 3 & -7 \\ -2 & 5 \end{bmatrix} = \begin{bmatrix} 226 & -525 \\ 90 & -209 \end{bmatrix}$

2.  $P = \begin{bmatrix} 2 & -3 \\ -3 & 5 \end{bmatrix}$ ,  $D = \begin{bmatrix} 1 & 0 \\ 0 & 1/2 \end{bmatrix}$ ,  $A = PDP^{-1}$ , and  $A^k = PD^kP^{-1}$ . We compute

$P^{-1} = \begin{bmatrix} 5 & 3 \\ 3 & 2 \end{bmatrix}$ ,  $D^k = \begin{bmatrix} 1 & 0 \\ 0 & 1/16 \end{bmatrix}$ , and  $A^k = \begin{bmatrix} 2 & -3 \\ -3 & 5 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 1/16 \end{bmatrix} \begin{bmatrix} 5 & 3 \\ 3 & 2 \end{bmatrix} = \frac{1}{16} \begin{bmatrix} 151 & 90 \\ -225 & -134 \end{bmatrix}$

3.  $A^k = PD^kP^{-1} = \begin{bmatrix} 1 & 0 \\ 3 & 1 \end{bmatrix} \begin{bmatrix} a^k & 0 \\ 0 & b^k \end{bmatrix} \begin{bmatrix} 1 & 0 \\ -3 & 1 \end{bmatrix} = \begin{bmatrix} a^k & 0 \\ 3a^k - 3b^k & b^k \end{bmatrix}$

4.  $A^k = PD^kP^{-1} = \begin{bmatrix} 3 & 4 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 2^k & 0 \\ 0 & 1^k \end{bmatrix} \begin{bmatrix} -1 & 4 \\ 1 & -3 \end{bmatrix} = \begin{bmatrix} 4 - 3 \cdot 2^k & 12 \cdot 2^k - 12 \\ 1 - 2^k & 4 \cdot 2^k - 3 \end{bmatrix}$

5. By the Diagonalization Theorem, eigenvectors form the columns of the left factor, and they correspond respectively to the eigenvalues on the diagonal of the middle factor.

$$\lambda = 5: \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}; \lambda = 1: \begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix}, \begin{bmatrix} 2 \\ -1 \\ 0 \end{bmatrix}$$

6. As in Exercise 5, inspection of the factorization gives:

$$\lambda = 4: \begin{bmatrix} -1 \\ 2 \\ 0 \end{bmatrix}; \lambda = 5: \begin{bmatrix} -2 \\ 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$$

7. Since  $A$  is triangular, its eigenvalues are obviously  $\pm 1$ .

For  $\lambda = 1$ :  $A - I = \begin{bmatrix} 0 & 0 \\ 6 & -2 \end{bmatrix}$ . The equation  $(A - I)\mathbf{x} = \mathbf{0}$  amounts to  $6x_1 - 2x_2 = 0$ , so  $x_1 = (1/3)x_2$  with

$x_2$  free. The general solution is  $x_2 \begin{bmatrix} 1/3 \\ 1 \end{bmatrix}$ , and a nice basis vector for the eigenspace is  $\mathbf{v}_1 = \begin{bmatrix} 1 \\ 3 \end{bmatrix}$ .

For  $\lambda = -1$ :  $A + I = \begin{bmatrix} 2 & 0 \\ 6 & 0 \end{bmatrix}$ . The equation  $(A + I)\mathbf{x} = \mathbf{0}$  amounts to  $2x_1 = 0$ , so  $x_1 = 0$  with  $x_2$  free.

The general solution is  $x_2 \begin{bmatrix} 0 \\ 1 \end{bmatrix}$ , and a basis vector for the eigenspace is  $\mathbf{v}_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$ .

From  $\mathbf{v}_1$  and  $\mathbf{v}_2$  construct  $P = [\mathbf{v}_1 \ \mathbf{v}_2] = \begin{bmatrix} 1 & 0 \\ 3 & 1 \end{bmatrix}$ . Then set  $D = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}$ , where the eigenvalues  $\pm 1$  correspond to  $\mathbf{v}_1$  and  $\mathbf{v}_2$  respectively.

8. Since  $A$  is triangular, its only eigenvalue is obviously 5.

For  $\lambda = 5$ :  $A - 5I = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$ . The equation  $(A - 5I)\mathbf{x} = \mathbf{0}$  amounts to  $x_2 = 0$ , so  $x_2 = 0$  with  $x_1$  free. The general solution is  $x_1 \begin{bmatrix} 1 \\ 0 \end{bmatrix}$ . Since we cannot generate an eigenvector basis for  $\mathbb{R}^2$ ,  $A$  is not diagonalizable.

9. To find the eigenvalues of  $A$ , compute its characteristic polynomial:

$$\det(A - \lambda I) = \det \begin{bmatrix} 3 - \lambda & -1 \\ 1 & 5 - \lambda \end{bmatrix} = (3 - \lambda)(5 - \lambda) - (-1)(1) = \lambda^2 - 8\lambda + 16 = (\lambda - 4)^2$$

Thus the only eigenvalue of  $A$  is 4.

For  $\lambda = 4$ :  $A - 4I = \begin{bmatrix} -1 & -1 \\ 1 & 1 \end{bmatrix}$ . The equation  $(A - 4I)\mathbf{x} = \mathbf{0}$  amounts to  $x_1 + x_2 = 0$ , so  $x_1 = -x_2$  with  $x_2$  free. The general solution is  $x_2 \begin{bmatrix} -1 \\ 1 \end{bmatrix}$ . Since we cannot generate an eigenvector basis for  $\mathbb{R}^2$ ,  $A$  is not diagonalizable.

10. To find the eigenvalues of  $A$ , compute its characteristic polynomial:

$$\det(A - \lambda I) = \det \begin{bmatrix} 2 - \lambda & 3 \\ 4 & 1 - \lambda \end{bmatrix} = (2 - \lambda)(1 - \lambda) - (3)(4) = \lambda^2 - 3\lambda - 10 = (\lambda - 5)(\lambda + 2)$$

Thus the eigenvalues of  $A$  are 5 and  $-2$ .

For  $\lambda = 5$ :  $A - 5I = \begin{bmatrix} -3 & 3 \\ 4 & -4 \end{bmatrix}$ . The equation  $(A - 5I)\mathbf{x} = \mathbf{0}$  amounts to  $x_1 - x_2 = 0$ , so  $x_1 = x_2$  with  $x_2$  free. The general solution is  $x_2 \begin{bmatrix} 1 \\ 1 \end{bmatrix}$ , and a basis vector for the eigenspace is  $\mathbf{v}_1 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$ .

For  $\lambda = -2$ :  $A + 2I = \begin{bmatrix} 4 & 3 \\ 4 & 3 \end{bmatrix}$ . The equation  $(A + 2I)\mathbf{x} = \mathbf{0}$  amounts to  $4x_1 + 3x_2 = 0$ , so  $x_1 = (-3/4)x_2$  with  $x_2$  free. The general solution is  $x_2 \begin{bmatrix} -3/4 \\ 1 \end{bmatrix}$ , and a nice basis vector for the eigenspace is  $\mathbf{v}_2 = \begin{bmatrix} -3 \\ 4 \end{bmatrix}$ .

From  $\mathbf{v}_1$  and  $\mathbf{v}_2$  construct  $P = [\mathbf{v}_1 \quad \mathbf{v}_2] = \begin{bmatrix} 1 & -3 \\ 1 & 4 \end{bmatrix}$ . Then set  $D = \begin{bmatrix} 5 & 0 \\ 0 & -2 \end{bmatrix}$ , where the eigenvalues in  $D$  correspond to  $\mathbf{v}_1$  and  $\mathbf{v}_2$  respectively.

11. The eigenvalues of  $A$  are given to be 1, 2, and 3.

For  $\lambda = 3$ :  $A - 3I = \begin{bmatrix} -4 & 4 & -2 \\ -3 & 1 & 0 \\ -3 & 1 & 0 \end{bmatrix}$ , and row reducing  $[A - 3I \quad \mathbf{0}]$  yields  $\begin{bmatrix} 1 & 0 & -1/4 & 0 \\ 0 & 1 & -3/4 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$ . The

general solution is  $x_3 \begin{bmatrix} 1/4 \\ 3/4 \\ 1 \end{bmatrix}$ , and a nice basis vector for the eigenspace is  $\mathbf{v}_1 = \begin{bmatrix} 1 \\ 3 \\ 4 \end{bmatrix}$ .

For  $\lambda = 2$ :  $A - 2I = \begin{bmatrix} -3 & 4 & -2 \\ -3 & 2 & 0 \\ -3 & 1 & 1 \end{bmatrix}$ , and row reducing  $[A - 2I \quad \mathbf{0}]$  yields  $\begin{bmatrix} 1 & 0 & -2/3 & 0 \\ 0 & 1 & -1 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$ . The

general solution is  $x_3 \begin{bmatrix} 2/3 \\ 1 \\ 1 \end{bmatrix}$ , and a nice basis vector for the eigenspace is  $\mathbf{v}_2 = \begin{bmatrix} 2 \\ 3 \\ 3 \end{bmatrix}$ .

For  $\lambda = 1$ :  $A - I = \begin{bmatrix} -2 & 4 & -2 \\ -3 & 3 & 0 \\ -3 & 1 & 2 \end{bmatrix}$ , and row reducing  $[A - I \quad \mathbf{0}]$  yields  $\begin{bmatrix} 1 & 0 & -1 & 0 \\ 0 & 1 & -1 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$ . The general

solution is  $x_3 \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$ , and a basis vector for the eigenspace is  $\mathbf{v}_1 = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$ .

From  $\mathbf{v}_1, \mathbf{v}_2$  and  $\mathbf{v}_3$  construct  $P = [\mathbf{v}_1 \quad \mathbf{v}_2 \quad \mathbf{v}_3] = \begin{bmatrix} 1 & 2 & 1 \\ 3 & 3 & 1 \\ 4 & 3 & 1 \end{bmatrix}$ . Then set  $D = \begin{bmatrix} 3 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 1 \end{bmatrix}$ , where the eigenvalues in  $D$  correspond to  $\mathbf{v}_1, \mathbf{v}_2$  and  $\mathbf{v}_3$  respectively.

12. The eigenvalues of  $A$  are given to be 2 and 8.

For  $\lambda = 8$ :  $A - 8I = \begin{bmatrix} -4 & 2 & 2 \\ 2 & -4 & 2 \\ 2 & 2 & -4 \end{bmatrix}$ , and row reducing  $[A - 8I \quad \mathbf{0}]$  yields  $\begin{bmatrix} 1 & 0 & -1 & 0 \\ 0 & 1 & -1 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$ . The

general solution is  $x_3 \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$ , and a basis vector for the eigenspace is  $\mathbf{v}_1 = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$ .

For  $\lambda = 2$ :  $A - 2I = \begin{bmatrix} 2 & 2 & 2 \\ 2 & 2 & 2 \\ 2 & 2 & 2 \end{bmatrix}$ , and row reducing  $[A - 2I \quad \mathbf{0}]$  yields  $\begin{bmatrix} 1 & 1 & 1 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$ . The general

solution is  $x_2 \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix} + x_3 \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}$ , and a basis for the eigenspace is  $(\mathbf{v}_2, \mathbf{v}_3) = \left\{ \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix} \right\}$ .

From  $\mathbf{v}_1, \mathbf{v}_2$  and  $\mathbf{v}_3$  construct  $P = [\mathbf{v}_1 \quad \mathbf{v}_2 \quad \mathbf{v}_3] = \begin{bmatrix} 1 & -1 & -1 \\ 1 & 1 & 0 \\ 1 & 0 & 1 \end{bmatrix}$ . Then set  $D = \begin{bmatrix} 8 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 2 \end{bmatrix}$ , where

eigenvalues in  $D$  correspond to  $\mathbf{v}_1, \mathbf{v}_2$  and  $\mathbf{v}_3$  respectively.

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

$$\begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \text{ The}$$

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

answer differs from  
new order of the

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

$$\begin{bmatrix} 2 \\ 0 \\ 0 \\ 1 \end{bmatrix}$$

$$\begin{bmatrix} 0 & 0 \\ 0 & 0 \\ 0 & 0 \\ 0 & 0 \end{bmatrix} \text{ The}$$

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

From  $\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3$  and  $\mathbf{v}_4$  construct  $P = [\mathbf{v}_1 \ \mathbf{v}_2 \ \mathbf{v}_3 \ \mathbf{v}_4] = \begin{bmatrix} 0 & 2 & 0 & 0 \\ 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 \end{bmatrix}$ . Then set  $D = \begin{bmatrix} 4 & 0 & 0 & 0 \\ 0 & 4 & 0 & 0 \\ 0 & 0 & 2 & 0 \\ 0 & 0 & 0 & 2 \end{bmatrix}$  where the eigenvalues in  $D$  correspond to  $\mathbf{v}_1, \mathbf{v}_2$  and  $\mathbf{v}_3$  respectively.

21. a. False. The symbol  $D$  does not automatically denote a diagonal matrix.  
 b. True. See the remark after the statement of the Diagonalization Theorem.  
 c. False. The  $3 \times 3$  matrix in Example 4 has 3 eigenvalues, counting multiplicities, but it is not diagonalizable.  
 d. False. Invertibility depends on 0 not being an eigenvalue. (See the Invertible Matrix Theorem.) A diagonalizable matrix may or may not have 0 as an eigenvalue. See Examples 3 and 5 for both possibilities.
22. a. False. The  $n$  eigenvectors must be linearly independent. See the Diagonalization Theorem.  
 b. False. The matrix in Example 3 is diagonalizable, but it has only 2 distinct eigenvalues. (The statement given is the *converse* of Theorem 6.)  
 c. True. This follows from  $AP = PD$  and formulas (1) and (2) in the proof of the Diagonalization Theorem.  
 d. False. See Example 4. The matrix there is invertible because 0 is not an eigenvalue, but the matrix is not diagonalizable.
23.  $A$  is diagonalizable because you know that five linearly independent eigenvectors exist: three in the three-dimensional eigenspace and two in the two-dimensional eigenspace. Theorem 7 guarantees that the set of all five eigenvectors is linearly independent.
24. No, by Theorem 7(b). Here is an explanation that does not appeal to Theorem 7: Let  $\mathbf{v}_1$  and  $\mathbf{v}_2$  be eigenvectors that span the two one-dimensional eigenspaces. If  $\mathbf{v}$  is any other eigenvector, then it belongs to one of the eigenspaces and hence is a multiple of either  $\mathbf{v}_1$  or  $\mathbf{v}_2$ . So there cannot exist three linearly independent eigenvectors. By the Diagonalization Theorem,  $A$  cannot be diagonalizable.
25. Let  $\{\mathbf{v}_1\}$  be a basis for the one-dimensional eigenspace, let  $\mathbf{v}_2$  and  $\mathbf{v}_3$  form a basis for the two-dimensional eigenspace, and let  $\mathbf{v}_4$  be any eigenvector in the remaining eigenspace. By Theorem 7,  $\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3, \mathbf{v}_4\}$  is linearly independent. Since  $A$  is  $4 \times 4$ , the Diagonalization Theorem shows that  $A$  is diagonalizable.
26. Yes, if the third eigenspace is only one-dimensional. In this case, the sum of the dimensions of the eigenspaces will be six, whereas the matrix is  $7 \times 7$ . See Theorem 7(b). An argument similar to that for Exercise 24 can also be given.
27. If  $A$  is diagonalizable, then  $A = PDP^{-1}$  for some invertible  $P$  and diagonal  $D$ . Since  $A$  is invertible, 0 is not an eigenvalue of  $A$ . So the diagonal entries in  $D$  (which are eigenvalues of  $A$ ) are not zero, and  $D$  is invertible. By the theorem on the inverse of a product,
- $$A^{-1} = (PDP^{-1})^{-1} = (P^{-1})^{-1} D^{-1} P^{-1} = PD^{-1}P^{-1}$$
- Since  $D^{-1}$  is obviously diagonal,  $A^{-1}$  is diagonalizable.

28. If  $A$  has  $n$  linearly independent eigenvectors, then by the Diagonalization Theorem,  $A = PDP^{-1}$  for some invertible  $P$  and diagonal  $D$ . Using properties of transposes,

$$\begin{aligned} A^T &= (PDP^{-1})^T = (P^{-1})^T D^T P^T \\ &= (P^T)^{-1} D P^T = QDQ^{-1} \end{aligned}$$

where  $Q = (P^T)^{-1}$ . Thus  $A^T$  is diagonalizable. By the Diagonalization Theorem, the columns of  $Q$  are  $n$  linearly independent eigenvectors of  $A^T$ .

29. The diagonal entries in  $D_1$  are reversed from those in  $D$ . So interchange the (eigenvector) columns of  $P$  to make them correspond properly to the eigenvalues in  $D_1$ . In this case,

$$P_1 = \begin{bmatrix} 1 & 1 \\ -2 & -1 \end{bmatrix} \text{ and } D_1 = \begin{bmatrix} 3 & 0 \\ 0 & 5 \end{bmatrix}$$

Although the first column of  $P$  must be an eigenvector corresponding to the eigenvalue 3, there is nothing to prevent us from selecting some multiple of  $\begin{bmatrix} 1 \\ -2 \end{bmatrix}$ , say  $\begin{bmatrix} -3 \\ 6 \end{bmatrix}$ , and letting  $P_2 = \begin{bmatrix} -3 & 1 \\ 6 & -1 \end{bmatrix}$ . We now have three different factorizations or "diagonalizations" of  $A$ :

$$A = PDP^{-1} = P_1 D_1 P_1^{-1} = P_2 D_1 P_2^{-1}$$

30. A nonzero multiple of an eigenvector is another eigenvector. To produce  $P_2$ , simply multiply one or both columns of  $P$  by a nonzero scalar unequal to 1.
31. For a  $2 \times 2$  matrix  $A$  to be invertible, its eigenvalues must be nonzero. A first attempt at a construction might be something such as  $\begin{bmatrix} 2 & 3 \\ 0 & 4 \end{bmatrix}$ , whose eigenvalues are 2 and 4. Unfortunately, a  $2 \times 2$  matrix with two distinct eigenvalues is diagonalizable (Theorem 6). So, adjust the construction to  $\begin{bmatrix} 2 & 3 \\ 0 & 2 \end{bmatrix}$ , which works. In fact, any matrix of the form  $\begin{bmatrix} a & b \\ 0 & a \end{bmatrix}$  has the desired properties when  $a$  and  $b$  are nonzero. The eigenspace for the eigenvalue  $a$  is one-dimensional, as a simple calculation shows, and there is no other eigenvalue to produce a second eigenvector.
32. Any  $2 \times 2$  matrix with two distinct eigenvalues is diagonalizable, by Theorem 6. If one of those eigenvalues is zero, then the matrix will not be invertible. Any matrix of the form  $\begin{bmatrix} a & b \\ 0 & 0 \end{bmatrix}$  has the desired properties when  $a$  and  $b$  are nonzero. The number  $a$  must be nonzero to make the matrix diagonalizable;  $b$  must be nonzero to make the matrix not diagonal. Other solutions are  $\begin{bmatrix} 0 & 0 \\ a & b \end{bmatrix}$

$$\text{and } \begin{bmatrix} 0 & a \\ 0 & b \end{bmatrix}.$$

3.  $A = \begin{bmatrix} 1 & 5 \\ -2 & 3 \end{bmatrix}$ . The characteristic polynomial is  $\lambda^2 - 4\lambda + 13$ , so the eigenvalues of  $A$  are

$$\lambda = \frac{4 \pm \sqrt{-36}}{2} = 2 \pm 3i.$$

For  $\lambda = 2 + 3i$ :  $A - (2 + 3i)I = \begin{bmatrix} -1 - 3i & 5 \\ -2 & 1 - 3i \end{bmatrix}$ . The equation  $(A - (2 + 3i)I)\mathbf{x} = \mathbf{0}$  amounts to  $-2x_1 + (1 - 3i)x_2 = 0$ , so  $x_1 = \frac{1 - 3i}{2}x_2$  with  $x_2$  free. A nice basis vector for the eigenspace is thus

$$\mathbf{v}_1 = \begin{bmatrix} 1 - 3i \\ 2 \end{bmatrix}.$$

For  $\lambda = 2 - 3i$ : A basis vector for the eigenspace is  $\mathbf{v}_2 = \bar{\mathbf{v}}_1 = \begin{bmatrix} 1 + 3i \\ 2 \end{bmatrix}$ .

4.  $A = \begin{bmatrix} 5 & -2 \\ 1 & 3 \end{bmatrix}$ . The characteristic polynomial is  $\lambda^2 - 8\lambda + 17$ , so the eigenvalues of  $A$  are

$$\lambda = \frac{8 \pm \sqrt{-4}}{2} = 4 \pm i.$$

For  $\lambda = 4 + i$ :  $A - (4 + i)I = \begin{bmatrix} 1 - i & -2 \\ 1 & -1 - i \end{bmatrix}$ . The equation  $(A - (4 + i)I)\mathbf{x} = \mathbf{0}$  amounts to  $x_1 + (-1 - i)x_2 = 0$ , so  $x_1 = (1 + i)x_2$  with  $x_2$  free. A basis vector for the eigenspace is thus

$$\mathbf{v}_1 = \begin{bmatrix} 1 + i \\ 1 \end{bmatrix}.$$

For  $\lambda = 4 - i$ : A basis vector for the eigenspace is  $\mathbf{v}_2 = \bar{\mathbf{v}}_1 = \begin{bmatrix} 1 - i \\ 1 \end{bmatrix}$ .

5.  $A = \begin{bmatrix} 0 & 1 \\ -8 & 4 \end{bmatrix}$ . The characteristic polynomial is  $\lambda^2 - 4\lambda + 8$ , so the eigenvalues of  $A$  are

$$\lambda = \frac{4 \pm \sqrt{-16}}{2} = 2 \pm 2i.$$

For  $\lambda = 2 + 2i$ :  $A - (2 + 2i)I = \begin{bmatrix} -2 - 2i & 1 \\ -8 & 2 - 2i \end{bmatrix}$ . The equation  $(A - (2 + 2i)I)\mathbf{x} = \mathbf{0}$  amounts to  $(-2 - 2i)x_1 + x_2 = 0$ , so  $x_2 = (2 + 2i)x_1$  with  $x_1$  free. A basis vector for the eigenspace is thus

$$\mathbf{v}_1 = \begin{bmatrix} 1 \\ 2 + 2i \end{bmatrix}.$$

For  $\lambda = 2 - 2i$ : A basis vector for the eigenspace is  $\mathbf{v}_2 = \bar{\mathbf{v}}_1 = \begin{bmatrix} 1 \\ 2 - 2i \end{bmatrix}$ .

6.  $A = \begin{bmatrix} 4 & 3 \\ -3 & 4 \end{bmatrix}$ . The characteristic polynomial is  $\lambda^2 - 8\lambda + 25$ , so the eigenvalues of  $A$  are

$$\lambda = \frac{8 \pm \sqrt{-36}}{2} = 4 \pm 3i.$$



For  $\lambda = 4 + 3i$ :  $A - (4 + 3i)I = \begin{bmatrix} -3i & 3 \\ -3 & -3i \end{bmatrix}$ . The equation  $(A - (4 + 3i)I)\mathbf{x} = \mathbf{0}$  amounts to  $x_1 + ix_2 = 0$ , so

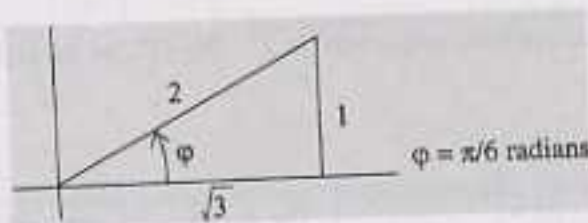
$x_1 = -ix_2$  with  $x_2$  free. A basis vector for the eigenspace is thus  $\mathbf{v}_1 = \begin{bmatrix} -i \\ 1 \end{bmatrix}$ .

For  $\lambda = 4 - 3i$ : A basis vector for the eigenspace is  $\mathbf{v}_2 = \bar{\mathbf{v}}_1 = \begin{bmatrix} i \\ 1 \end{bmatrix}$ .

7.  $A = \begin{bmatrix} \sqrt{3} & -1 \\ 1 & \sqrt{3} \end{bmatrix}$ . From Example 6, the eigenvalues are  $\sqrt{3} \pm i$ . The scale factor for the transformation

$\mathbf{x} \mapsto A\mathbf{x}$  is  $r = |\lambda| = \sqrt{(\sqrt{3})^2 + 1^2} = 2$ . For the angle of rotation, plot the point  $(a, b) = (\sqrt{3}, 1)$  in the  $xy$ -plane and use trigonometry:

$$\varphi = \arctan(b/a) = \arctan(1/\sqrt{3}) = \pi/6 \text{ radians.}$$



**Note:** Your students will want to know whether you permit them on an exam to omit calculations for a matrix of the form  $\begin{bmatrix} a & -b \\ b & a \end{bmatrix}$  and simply write the eigenvalues  $a \pm bi$ . A similar question may arise about the corresponding eigenvectors,  $\begin{bmatrix} 1 \\ -i \end{bmatrix}$  and  $\begin{bmatrix} 1 \\ i \end{bmatrix}$ , which are announced in the Practice Problem. Students may have trouble keeping track of the correspondence between eigenvalues and eigenvectors.

8.  $A = \begin{bmatrix} \sqrt{3} & 3 \\ -3 & \sqrt{3} \end{bmatrix}$ . From Example 6, the eigenvalues are  $\sqrt{3} \pm 3i$ . The scale factor for the transformation

$\mathbf{x} \mapsto A\mathbf{x}$  is  $r = |\lambda| = \sqrt{(\sqrt{3})^2 + 3^2} = 2\sqrt{3}$ . From trigonometry, the angle of rotation  $\varphi$  is  $\arctan(b/a) = \arctan(-3/\sqrt{3}) = -\pi/3$  radians.

9.  $A = \begin{bmatrix} -\sqrt{3}/2 & 1/2 \\ -1/2 & -\sqrt{3}/2 \end{bmatrix}$ . From Example 6, the eigenvalues are  $-\sqrt{3}/2 \pm (1/2)i$ . The scale factor for the

transformation  $\mathbf{x} \mapsto A\mathbf{x}$  is  $r = |\lambda| = \sqrt{(-\sqrt{3}/2)^2 + (1/2)^2} = 1$ . From trigonometry, the angle of rotation  $\varphi$  is  $\arctan(b/a) = \arctan((-1/2)/(-\sqrt{3}/2)) = -5\pi/6$  radians.

15.  $A = \begin{bmatrix} 1 & 5 \\ -2 & 3 \end{bmatrix}$ . From Exercise 3, the eigenvalues of  $A$  are  $\lambda = 2 \pm 3i$ , and the eigenvector

$\mathbf{v} = \begin{bmatrix} 1+3i \\ 2 \end{bmatrix}$  corresponds to  $\lambda = 2 - 3i$ . By Theorem 9,  $P = [\operatorname{Re} \mathbf{v} \quad \operatorname{Im} \mathbf{v}] = \begin{bmatrix} 1 & 3 \\ 2 & 0 \end{bmatrix}$  and

$$C = P^{-1}AP = \frac{1}{6} \begin{bmatrix} 0 & -3 \\ -2 & 1 \end{bmatrix} \begin{bmatrix} 1 & 5 \\ -2 & 3 \end{bmatrix} \begin{bmatrix} 1 & 3 \\ 2 & 0 \end{bmatrix} = \begin{bmatrix} 2 & -3 \\ 3 & 2 \end{bmatrix}$$

16.  $A = \begin{bmatrix} 5 & -2 \\ 1 & 3 \end{bmatrix}$ . From Exercise 4, the eigenvalues of  $A$  are  $\lambda = 4 \pm i$ , and the eigenvector

$\mathbf{v} = \begin{bmatrix} 1-i \\ 1 \end{bmatrix}$  corresponds to  $\lambda = 4 - i$ . By Theorem 9,  $P = [\operatorname{Re} \mathbf{v} \quad \operatorname{Im} \mathbf{v}] = \begin{bmatrix} 1 & -1 \\ 1 & 0 \end{bmatrix}$  and

$$C = P^{-1}AP = \begin{bmatrix} 0 & 1 \\ -1 & 1 \end{bmatrix} \begin{bmatrix} 5 & -2 \\ 1 & 3 \end{bmatrix} \begin{bmatrix} 1 & -1 \\ 1 & 0 \end{bmatrix} = \begin{bmatrix} 4 & -1 \\ 1 & 4 \end{bmatrix}$$

17.  $A = \begin{bmatrix} 1 & -8 \\ 4 & -2.2 \end{bmatrix}$ . The characteristic polynomial is  $\lambda^2 + 1.2\lambda + 1$ , so the eigenvalues of  $A$  are  $\lambda = -.6 \pm .8i$ .

To find an eigenvector corresponding to  $-.6 - .8i$ , we compute

$$A - (-.6 - .8i)I = \begin{bmatrix} 1.6 + .8i & -8 \\ 4 & -1.6 + .8i \end{bmatrix}$$

The equation  $(A - (-.6 - .8i)I)\mathbf{x} = \mathbf{0}$  amounts to  $.4x_1 + (-1.6 + .8i)x_2 = 0$ , so  $x_1 = ((2-i)/5)x_2$

with  $x_2$  free. A nice eigenvector corresponding to  $-.6 - .8i$  is thus  $\mathbf{v} = \begin{bmatrix} 2-i \\ 5 \end{bmatrix}$ . By Theorem 9,

$$P = [\operatorname{Re} \mathbf{v} \quad \operatorname{Im} \mathbf{v}] = \begin{bmatrix} 2 & -1 \\ 5 & 0 \end{bmatrix} \text{ and } C = P^{-1}AP = \frac{1}{5} \begin{bmatrix} 0 & 1 \\ -5 & 1 \end{bmatrix} \begin{bmatrix} 1 & -8 \\ 4 & -2.2 \end{bmatrix} \begin{bmatrix} 2 & -1 \\ 5 & 0 \end{bmatrix} = \begin{bmatrix} -.6 & -.8 \\ .8 & -.6 \end{bmatrix}$$

18.  $A = \begin{bmatrix} 1 & -1 \\ 4 & .6 \end{bmatrix}$ . The characteristic polynomial is  $\lambda^2 - 1.6\lambda + 1$ , so the eigenvalues of  $A$  are  $\lambda = .8 \pm .6i$ . To

find an eigenvector corresponding to  $.8 - .6i$ , we compute

$$A - (.8 - .6i)I = \begin{bmatrix} .2 + .6i & -1 \\ 4 & -.2 + .6i \end{bmatrix}$$

The equation  $(A - (.8 - .6i)I)\mathbf{x} = \mathbf{0}$  amounts to  $.4x_1 + (-.2 + .6i)x_2 = 0$ , so  $x_1 = ((1-3i)/2)x_2$  with  $x_2$  free.

A nice eigenvector corresponding to  $.8 - .6i$  is thus  $\mathbf{v} = \begin{bmatrix} 1-3i \\ 2 \end{bmatrix}$ . By Theorem 9,

$$P = [\operatorname{Re} \mathbf{v} \quad \operatorname{Im} \mathbf{v}] = \begin{bmatrix} 1 & -3 \\ 2 & 0 \end{bmatrix} \text{ and } C = P^{-1}AP = \frac{1}{6} \begin{bmatrix} 0 & 3 \\ -2 & 1 \end{bmatrix} \begin{bmatrix} 1 & -1 \\ 4 & .6 \end{bmatrix} \begin{bmatrix} 1 & -3 \\ 2 & 0 \end{bmatrix} = \begin{bmatrix} .8 & -.6 \\ .6 & .8 \end{bmatrix}$$

19.  $A = \begin{bmatrix} 1.52 & -.7 \\ .56 & .4 \end{bmatrix}$ . The characteristic polynomial is  $\lambda^2 - 1.92\lambda + 1$ , so the eigenvalues of  $A$  are  $\lambda = .96 \pm .28i$ . To find an eigenvector corresponding to  $.96 - .28i$ , we compute

$$A - (.96 - .28i)I = \begin{bmatrix} .56 + .28i & -.7 \\ .56 & -.56 + .28i \end{bmatrix}$$

The equation  $(A - (.96 - .28i)I)\mathbf{x} = \mathbf{0}$  amounts to  $.56x_1 + (-.56 + .28i)x_2 = 0$ , so  $x_1 = ((2-i)/2)x_2$  with  $x_2$  free. A nice eigenvector corresponding to  $.96 - .28i$  is thus  $\mathbf{v} = \begin{bmatrix} 2-i \\ 2 \end{bmatrix}$ . By Theorem 9,

$$P = [\operatorname{Re} \mathbf{v} \quad \operatorname{Im} \mathbf{v}] = \begin{bmatrix} 2 & -1 \\ 2 & 0 \end{bmatrix} \text{ and } C = P^{-1}AP = \frac{1}{2} \begin{bmatrix} 0 & 1 \\ -2 & 2 \end{bmatrix} \begin{bmatrix} 1.52 & -.7 \\ .56 & .4 \end{bmatrix} \begin{bmatrix} 2 & -1 \\ 2 & 0 \end{bmatrix} = \begin{bmatrix} .96 & -.28 \\ .28 & .96 \end{bmatrix}$$

20.  $A = \begin{bmatrix} -1.64 & -2.4 \\ 1.92 & 2.2 \end{bmatrix}$ . The characteristic polynomial is  $\lambda^2 - .56\lambda + 1$ , so the eigenvalues of  $A$  are  $\lambda = .28 \pm .96i$ . To find an eigenvector corresponding to  $.28 - .96i$ , we compute

$$A - (.28 - .96i)I = \begin{bmatrix} -1.92 + .96i & -2.4 \\ 1.92 & 1.92 + .96i \end{bmatrix}$$

The equation  $(A - (.28 - .96i)I)\mathbf{x} = \mathbf{0}$  amounts to  $1.92x_1 + (1.92 + .96i)x_2 = 0$ , so  $x_1 = ((-2-i)/2)x_2$  with  $x_2$  free. A nice eigenvector corresponding to  $.28 - .96i$  is thus  $\mathbf{v} = \begin{bmatrix} -2-i \\ 2 \end{bmatrix}$ . By Theorem 9,

$$P = [\operatorname{Re} \mathbf{v} \quad \operatorname{Im} \mathbf{v}] = \begin{bmatrix} -2 & -1 \\ 2 & 0 \end{bmatrix} \text{ and } C = P^{-1}AP = \frac{1}{2} \begin{bmatrix} 0 & 1 \\ -2 & -2 \end{bmatrix} \begin{bmatrix} -1.64 & -2.4 \\ 1.92 & 2.2 \end{bmatrix} \begin{bmatrix} -2 & -1 \\ 2 & 0 \end{bmatrix} = \begin{bmatrix} .28 & -.96 \\ .96 & .28 \end{bmatrix}$$

21. The first equation in (2) is  $(-.3 + .6i)x_1 - .6x_2 = 0$ . We solve this for  $x_2$  to find that  $x_2 = ((-.3 + .6i)/.6)x_1 = ((-1 + 2i)/2)x_1$ . Letting  $x_1 = 2$ , we find that  $\mathbf{y} = \begin{bmatrix} 2 \\ -1 + 2i \end{bmatrix}$  is an eigenvector for the matrix  $A$ . Since  $\mathbf{y} = \begin{bmatrix} 2 \\ -1 + 2i \end{bmatrix} = \frac{-1 + 2i}{5} \begin{bmatrix} -2 - 4i \\ 5 \end{bmatrix} = \frac{-1 + 2i}{5} \mathbf{v}_1$ , the vector  $\mathbf{y}$  is a complex multiple of the vector  $\mathbf{v}_1$  used in Example 2.

22. Since  $A(\mu \mathbf{x}) = \mu(A\mathbf{x}) = \mu(\lambda \mathbf{x}) = \lambda(\mu \mathbf{x})$ ,  $\mu \mathbf{x}$  is an eigenvector of  $A$ .

23. (a) properties of conjugates and the fact that  $\overline{\overline{\mathbf{x}}^T} = \mathbf{x}^T$   
 (b)  $\overline{A\mathbf{x}} = A\overline{\mathbf{x}}$  and  $A$  is real  
 (c)  $\mathbf{x}^T A\overline{\mathbf{x}}$  is a scalar and hence may be viewed as a  $1 \times 1$  matrix.  
 (d) properties of transposes  
 (e)  $A^T = A$  and the definition of  $q$

24.  $\overline{\mathbf{x}}^T A\mathbf{x} = \overline{\mathbf{x}}^T (\lambda \mathbf{x}) = \lambda \cdot \overline{\mathbf{x}}^T \mathbf{x}$  because  $\mathbf{x}$  is an eigenvector. It is easy to see that  $\overline{\mathbf{x}}^T \mathbf{x}$  is real (and positive) because  $\overline{z}$  is nonnegative for every complex number  $z$ . Since  $\overline{\mathbf{x}}^T A\mathbf{x}$  is real, by Exercise 23, so is  $\lambda$ . Next, write  $\mathbf{x} = \mathbf{u} + i\mathbf{v}$ , where  $\mathbf{u}$  and  $\mathbf{v}$  are real vectors. Then  $A\mathbf{x} = A(\mathbf{u} + i\mathbf{v}) = A\mathbf{u} + iA\mathbf{v}$  and  $\lambda \mathbf{x} = \lambda \mathbf{u} + i\lambda \mathbf{v}$

The real part of  $Ax$  is  $Au$  because the entries in  $A$ ,  $u$ , and  $v$  are all real. The real part of  $\lambda x$  is  $\lambda u$  because  $\lambda$  and the entries in  $u$  and  $v$  are real. Since  $Ax$  and  $\lambda x$  are equal, their real parts are equal, too. (Apply the corresponding statement about complex numbers to each entry of  $Ax$ .) Thus  $Au = \lambda u$ , which shows that the real part of  $x$  is an eigenvector of  $A$ .

25. Write  $x = \operatorname{Re} x + i(\operatorname{Im} x)$ , so that  $Ax = A(\operatorname{Re} x) + iA(\operatorname{Im} x)$ . Since  $A$  is real, so are  $A(\operatorname{Re} x)$  and  $A(\operatorname{Im} x)$ . Thus  $A(\operatorname{Re} x)$  is the real part of  $Ax$  and  $A(\operatorname{Im} x)$  is the imaginary part of  $Ax$ .

26. a. If  $\lambda = a - bi$ , then

$$\begin{aligned} Av &= \lambda v = (a - bi)(\operatorname{Re} v + i \operatorname{Im} v) \\ &= \underbrace{(a \operatorname{Re} v + b \operatorname{Im} v)}_{\operatorname{Re} Av} + i \underbrace{(a \operatorname{Im} v - b \operatorname{Re} v)}_{\operatorname{Im} Av} \end{aligned}$$

By Exercise 25,

$$A(\operatorname{Re} v) = \operatorname{Re} Av = a \operatorname{Re} v + b \operatorname{Im} v$$

$$A(\operatorname{Im} v) = \operatorname{Im} Av = -b \operatorname{Re} v + a \operatorname{Im} v$$

b. Let  $P = [\operatorname{Re} v \quad \operatorname{Im} v]$ . By (a),

$$A(\operatorname{Re} v) = P \begin{bmatrix} a \\ b \end{bmatrix}, \quad A(\operatorname{Im} v) = P \begin{bmatrix} -b \\ a \end{bmatrix}$$

So

$$\begin{aligned} AP &= [A(\operatorname{Re} v) \quad A(\operatorname{Im} v)] \\ &= \left[ P \begin{bmatrix} a \\ b \end{bmatrix} \quad P \begin{bmatrix} -b \\ a \end{bmatrix} \right] = P \begin{bmatrix} a & -b \\ b & a \end{bmatrix} = PC \end{aligned}$$

$$27. [M] \quad A = \begin{bmatrix} .7 & 1.1 & 2.0 & 1.7 \\ -2.0 & -4.0 & -8.6 & -7.4 \\ 0 & -5 & -1.0 & -1.0 \\ 1.0 & 2.8 & 6.0 & 5.3 \end{bmatrix}$$

$$\operatorname{ev} = \operatorname{eig}(A) = (.2 + .5i, .2 - .5i, .3 + .1i, .3 - .1i)$$

For  $\lambda = .2 - .5i$ , an eigenvector is

$$\operatorname{nulbasis}(A - \operatorname{ev}(2) * \operatorname{eye}(4)) =$$

$$0.5000 - 0.5000i$$

$$-2.0000 + 0.0000i$$

$$0.0000 - 0.0000i$$

$$1.0000$$

$$\text{so that } v_1 = \begin{bmatrix} .5 - .5i \\ -2 \\ 0 \\ 1 \end{bmatrix}$$

For  $\lambda = .3 - .1i$ , an eigenvector is

$$\operatorname{nulbasis}(A - \operatorname{ev}(4) * \operatorname{eye}(4)) =$$

$$-0.5000 - 0.0000i$$