

## 8.2

$$1. A = \begin{bmatrix} 6 & -\frac{7}{2} \\ -\frac{7}{2} & 8 \end{bmatrix}$$

$$2. A = \begin{bmatrix} 0 & \frac{1}{2} \\ \frac{1}{2} & 0 \end{bmatrix}$$

$$3. A = \begin{bmatrix} 3 & 0 & 3 \\ 0 & 4 & \frac{7}{2} \\ 3 & \frac{7}{2} & 5 \end{bmatrix}$$

4.  $A = \begin{bmatrix} 6 & 2 \\ 2 & 3 \end{bmatrix}$ , positive definite

5.  $A = \begin{bmatrix} 1 & 2 \\ 2 & 1 \end{bmatrix}$ , indefinite (since  $\det(A) < 0$ )

6.  $A = \begin{bmatrix} 2 & 3 \\ 3 & 4 \end{bmatrix}$ , indefinite (since  $\det(A) < 0$ )

7.  $A = \begin{bmatrix} 0 & 0 & 2 \\ 0 & 3 & 0 \\ 2 & 0 & 0 \end{bmatrix}$ , indefinite (eigenvalues 2, -2, 3)

8. If  $S^{-1}AS = D$  is diagonal, then  $S^{-1}A^2S = D^2$ , so that all eigenvalues of  $A^2$  are  $\geq 0$ . So  $A^2$  is positive semi-definite; it is positive definite if and only if  $A$  is invertible.

9. a.  $(A^2)^T = (A^T)^2 = (-A)^2 = A^2$ , so that  $A^2$  is symmetric.

b.  $q(\vec{x}) = \vec{x}^T A^2 \vec{x} = \vec{x}^T A A \vec{x} = -\vec{x}^T A^T A \vec{x} = -(A\vec{x}) \cdot (A\vec{x}) = -\|A\vec{x}\|^2 \leq 0$  for all  $\vec{x}$ , so that  $A^2$  is negative semi-definite.

c. If  $\vec{v}$  is a complex eigenvector of  $A$  with eigenvalue  $\lambda$ , then  $A^2\vec{v} = \lambda^2\vec{v}$ , and  $\lambda^2 \leq 0$ , by part b. Therefore,  $\lambda$  is *imaginary*, that is,  $\lambda = bi$  for a real  $b$ . Thus, the zero matrix is the only skew-symmetric matrix that is diagonalizable over  $\mathbb{R}$ .

10.  $L(\vec{x}) = (\vec{x} + \vec{v})^T A(\vec{x} + \vec{v}) - \vec{x}^T A \vec{x} - \vec{v}^T A \vec{v} = \vec{x}^T A \vec{x} + \vec{x}^T A \vec{v} + \vec{v}^T A \vec{x} + \vec{v}^T A \vec{v} - \vec{x}^T A \vec{x} - \vec{v}^T A \vec{v} = \vec{x}^T A \vec{v} + \vec{v}^T A \vec{x} = \vec{v}^T A \vec{x} + \vec{v}^T A \vec{x} = (2\vec{v}^T A)\vec{x}$ ,

↑

note that  $\vec{x}^T A \vec{v}$  is a scalar so that  $\vec{x}^T A \vec{v} = (\vec{x}^T A \vec{v})^T = \vec{v}^T A^T \vec{x} = \vec{v}^T A \vec{x}$  if  $A$  is symmetric.

So  $L$  is linear with matrix  $2\vec{v}^T A$ .

11. The eigenvalues of  $A^{-1}$  are the reciprocals of those of  $A$ , so that  $A$  and  $A^{-1}$  have the same definiteness.

12.  $\det(A)$  is the product of the two (real) eigenvalues.  $q$  is indefinite if and only if those have different signs, that is, their product is negative.

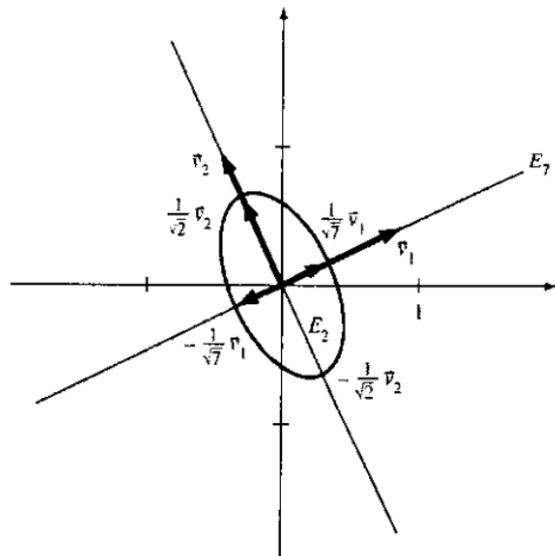
13.  $q(\vec{e}_i) = \vec{e}_i \cdot A\vec{e}_i = a_{ii} > 0$

14. If  $\det(A)$  is positive then both eigenvalues have the same sign, so that  $A$  is positive definite or negative definite. Since  $\vec{e}_1 \cdot A\vec{e}_1 = a > 0$ ,  $A$  is in fact positive definite.

15.  $A = \begin{bmatrix} 6 & 2 \\ 2 & 3 \end{bmatrix}$ ; eigenvalues  $\lambda_1 = 7$  and  $\lambda_2 = 2$

orthonormal eigenbasis  $\vec{v}_1 = \frac{1}{\sqrt{5}} \begin{bmatrix} 2 \\ 1 \end{bmatrix}$ ,  $\vec{v}_2 = \frac{1}{\sqrt{5}} \begin{bmatrix} -1 \\ 2 \end{bmatrix}$

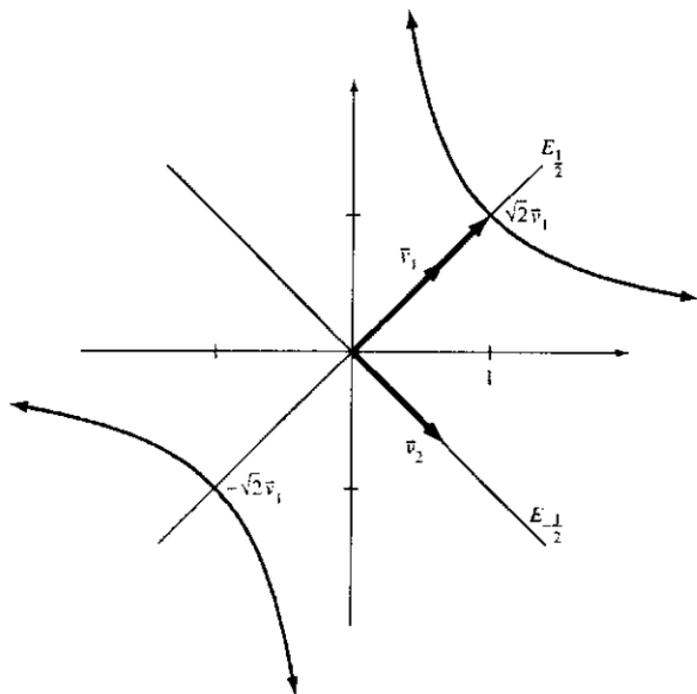
$\lambda_1 c_1^2 + \lambda_2 c_2^2 = 1$  or  $7c_1^2 + 2c_2^2 = 1$



16.  $A = \begin{bmatrix} 0 & \frac{1}{2} \\ \frac{1}{2} & 0 \end{bmatrix}$ ; eigenvalues  $\lambda_1 = \frac{1}{2}$ , and  $\lambda_2 = -\frac{1}{2}$

orthonormal eigenbasis  $\vec{v}_1 = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 1 \end{bmatrix}$  and  $\vec{v}_2 = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ -1 \end{bmatrix}$

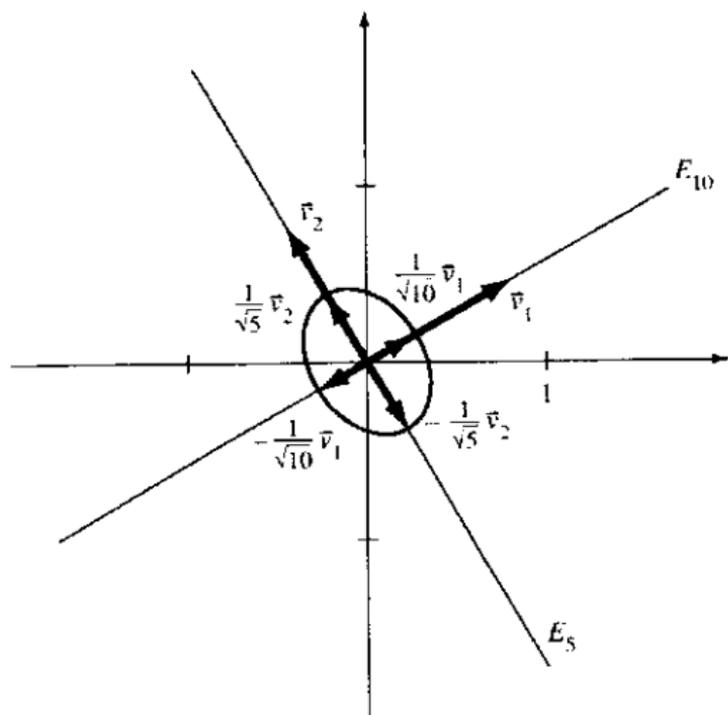
$$\frac{1}{2}c_1^2 - \frac{1}{2}c_2^2 = 1$$



18.  $A = \begin{bmatrix} 9 & -2 \\ -2 & 6 \end{bmatrix}$ , eigenvalues  $\lambda_1 = 10, \lambda_2 = 5$

orthonormal eigenbasis  $\vec{v}_1 = \frac{1}{\sqrt{5}} \begin{bmatrix} 2 \\ 1 \end{bmatrix}, \vec{v}_2 = \frac{1}{\sqrt{5}} \begin{bmatrix} -1 \\ 2 \end{bmatrix}$

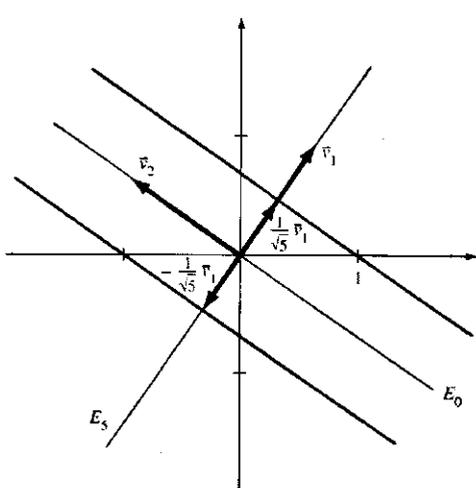
$10c_1^2 + 5c_2^2 = 1$  (ellipse)



19.  $A = \begin{bmatrix} 1 & 2 \\ 2 & 4 \end{bmatrix}$ ; eigenvalues  $\lambda_1 = 5, \lambda_2 = 0$

eigenvectors  $\vec{v}_1 = \frac{1}{\sqrt{5}} \begin{bmatrix} 1 \\ 2 \end{bmatrix}, \vec{v}_2 = \frac{1}{\sqrt{5}} \begin{bmatrix} -2 \\ 1 \end{bmatrix}$

$5c_1^2 = 1$  (a pair of lines)

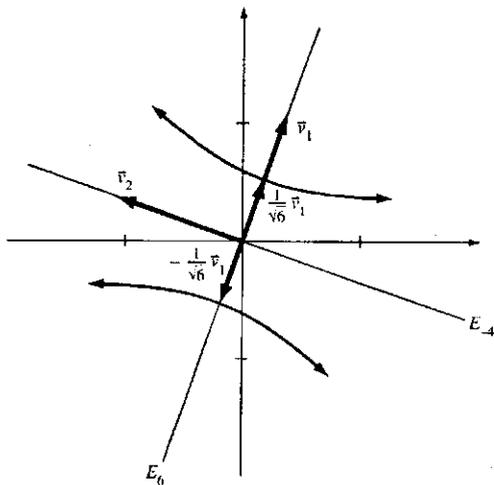


Note that  $(x_1^2 + 4x_1x_2 + 4x_2^2) = (x_1 + 2x_2)^2 = 1$ , so that  $x_1 + 2x_2 = \pm 1$ , and the two lines are  $x_2 = \frac{1-x_1}{2}$  and  $x_2 = \frac{-1-x_1}{2}$ .

20.  $A = \begin{bmatrix} -3 & 3 \\ 3 & 5 \end{bmatrix}$ ; eigenvalues  $\lambda_1 = 6$  and  $\lambda_2 = -4$

orthonormal eigenbasis  $\vec{v}_1 = \frac{1}{\sqrt{10}} \begin{bmatrix} 1 \\ 3 \end{bmatrix}$ ,  $\vec{v}_2 = \frac{1}{\sqrt{10}} \begin{bmatrix} -3 \\ 1 \end{bmatrix}$

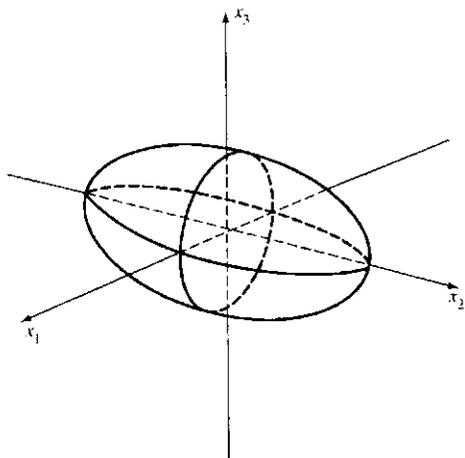
$6c_1^2 - 4c_2^2 = 1$  (hyperbola)



21. a. In each case, it is informative to think about the intersections with the three coordinate planes:  $x_1 - x_2$ ,  $x_1 - x_3$ , and  $x_2 - x_3$ .

For the surface  $x_1^2 + 4x_2^2 + 9x_3^2 = 1$ , all these intersections are *ellipses*, and the surface itself is an *ellipsoid*.

This surface is connected and bounded; the points closest to the origin are  $\pm \begin{bmatrix} 0 \\ 0 \\ \frac{1}{3} \end{bmatrix}$ , and those farthest  $\pm \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$ .

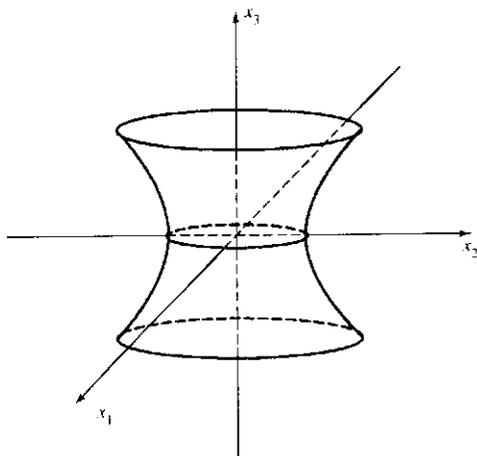


$$x_1^2 + 4x_2^2 + 9x_3^2 = 1 \text{ (not to scale)}$$

an *ellipsoid*

In the case of  $x_1^2 + 4x_2^2 - 9x_3^2 = 1$ , the intersection with the  $x_1 - x_2$  plane is an ellipse, and the two other intersections are hyperbolas. The surface is connected and not bounded; the points closest to

the origin are  $\pm \begin{bmatrix} 0 \\ 0 \\ \frac{1}{2} \end{bmatrix}$ .

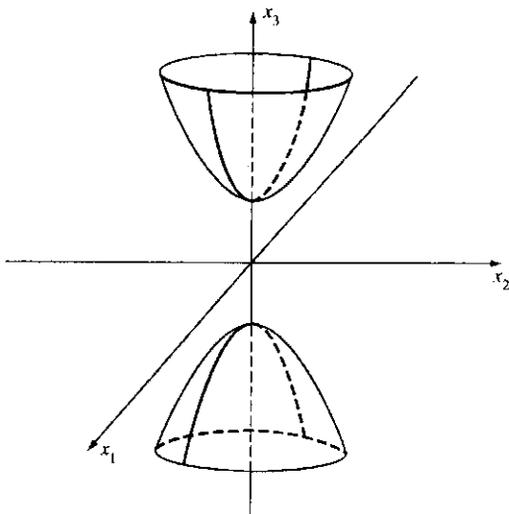


$$x_1^2 + 4x_2^2 - 9x_3^2 = 1 \text{ (not to scale)}$$

a *hyperboloid of one sheet*

In the case  $-x_1^2 - 4x_2^2 + 9x_3^2 = 1$ , the intersection with the  $x_1 - x_2$  plane is empty, and the two other intersections are hyperbolas. The surface consists of two pieces and is unbounded. The points closest

to the origin are  $\pm \begin{bmatrix} 0 \\ 0 \\ \frac{1}{3} \end{bmatrix}$ .



$-x_1^2 - 4x_2^2 + 9x_3^2 = 1$  (not to scale)  
a *hyperboloid of two sheets*

b.  $A = \begin{bmatrix} 1 & \frac{1}{2} & 1 \\ \frac{1}{2} & 2 & \frac{3}{2} \\ 1 & \frac{3}{2} & 3 \end{bmatrix}$  is positive definite, with three positive eigenvalues  $\lambda_1, \lambda_2, \lambda_3$ .

Surface is given by  $\lambda_1 c_1^2 + \lambda_2 c_2^2 + \lambda_3 c_3^2 = 1$  with respect to principal axes, an *ellipsoid*. To find points closest to and farthest from origin, use technology to find eigenvalues and eigenvectors:  
eigenvalues:  $\lambda_1 \approx 0.56, \lambda_2 \approx 4.44, \lambda_3 = 1$

unit eigenvectors:  $\vec{v}_1 \approx \begin{bmatrix} 0.86 \\ 0.19 \\ -0.47 \end{bmatrix}, \vec{v}_2 \approx \begin{bmatrix} 0.31 \\ 0.54 \\ 0.78 \end{bmatrix}, \vec{v}_3 = \frac{1}{\sqrt{6}} \begin{bmatrix} 1 \\ -2 \\ 1 \end{bmatrix}$

Equation:  $0.56c_1^2 + 4.44c_2^2 + c_3^2 = 1$

Farthest points when  $c_1 = \pm \frac{1}{\sqrt{0.56}}$  and  $c_2 = c_3 = 0$

Closest points when  $c_2 = \pm \frac{1}{\sqrt{4.44}}$  and  $c_1 = c_3 = 0$

Farthest points  $\approx \pm \frac{1}{\sqrt{0.56}} \begin{bmatrix} 0.86 \\ 0.19 \\ -0.47 \end{bmatrix} \approx \pm \begin{bmatrix} 1.15 \\ 0.26 \\ -0.63 \end{bmatrix}$

$$\text{Closest points} \approx \pm \frac{1}{\sqrt{4.44}} \begin{bmatrix} 0.31 \\ 0.54 \\ 0.78 \end{bmatrix} \approx \pm \begin{bmatrix} 0.15 \\ 0.26 \\ 0.37 \end{bmatrix}$$

$$22. A = \begin{bmatrix} -1 & 0 & 5 \\ 0 & 1 & 0 \\ 5 & 0 & -1 \end{bmatrix}; \text{ eigenvalues } \lambda_1 = 4, \lambda_2 = -6, \lambda_3 = 1$$

Equation with respect to principal axes:  $4c_1^2 - 6c_2^2 + c_3^2 = 1$ , a hyperboloid of one sheet (see solutions to 21a).

Closest to origin when  $c_1 = \pm \frac{1}{2}$ ,  $c_2 = c_3 = 0$ .

A unit eigenvector for eigenvalue 4 is  $\vec{v} = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}$ , so that the desired points are  $\pm \frac{1}{2} \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} \approx$   
 $\pm \begin{bmatrix} 0.35 \\ 0 \\ 0.35 \end{bmatrix}$ .