

## Section 6.2 Problems

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$QQ^T = I$  is analogous to  $\pm 1^2 = 1$

**1.**

$$A = \begin{bmatrix} a & 2 & 2 \\ 2 & a & 2 \\ 2 & 2 & a \end{bmatrix} \quad B = \begin{bmatrix} 1 & 2 & 4 \\ 2 & b & 8 \\ 4 & 8 & 7 \end{bmatrix}$$

$A$ :  $a > 2$ .  $B$ : no numbers

**2.**

$A$ : pivots are  $(2, 3/2, 0)$ , so no

$B$ : pivots are  $(2, 3/2, 1)$ , so yes

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

By inspection, it's obvious some pivots are 0.

**3.**

$$\begin{aligned} \det A &= 1(1 - b^2) - b(b + b^2) + -b(b^2 + b) \\ &= 1 - b^2 - 2(b^3 + b^2) \\ &= -2b^3 - 3b^2 + 1 \end{aligned}$$

Can't find an obvious value to make it negative.

**4.**

We can diagonalize since  $A$  is symmetric.

$$\begin{aligned} A &= S\Lambda S^{-1} \\ A^2 &= S\Lambda^2 S^{-1} \\ A^{-1} &= S\Lambda^{-1} S^{-1} \end{aligned}$$

5.

$$\begin{aligned}x^T(A+B)x &> 0 \\ x^T Ax + x^T Bx &> 0\end{aligned}$$

Both terms are greater than zero, so their sum is too.

6.

Some factorizations

$$A = \begin{bmatrix} 5 & 4 \\ 4 & 5 \end{bmatrix}$$

$$\begin{aligned}(L\sqrt{D})(\sqrt{D}L^T) &= \left( \begin{bmatrix} 1 & 0 \\ 4/5 & 1 \end{bmatrix} \begin{bmatrix} \sqrt{5} & 0 \\ 0 & 3/\sqrt{5} \end{bmatrix} \right) \left( \begin{bmatrix} \sqrt{5} & 0 \\ 0 & 3/\sqrt{5} \end{bmatrix} \begin{bmatrix} 1 & 4/5 \\ 0 & 1 \end{bmatrix} \right) \\ (Q\Lambda)(\Lambda Q^T) &= \left( \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{2} \\ 1/\sqrt{2} & -1/\sqrt{2} \end{bmatrix} \begin{bmatrix} 3 & 0 \\ 0 & 1 \end{bmatrix} \right) \left( \begin{bmatrix} 3 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{2} \\ 1/\sqrt{2} & -1/\sqrt{2} \end{bmatrix} \right) \\ (Q\sqrt{\Lambda}Q^T)^2 &= \left( \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{2} \\ 1/\sqrt{2} & -1/\sqrt{2} \end{bmatrix} \begin{bmatrix} 3 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{2} \\ 1/\sqrt{2} & -1/\sqrt{2} \end{bmatrix} \right)\end{aligned}$$

7.

As a square root of symmetric PD  $A$ ,  $R = Q\sqrt{\Lambda}Q$  has the same signs of eigenvalues.

$$A = \left( \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{2} \\ 1/\sqrt{2} & -1/\sqrt{2} \end{bmatrix} \begin{bmatrix} 4 & 0 \\ 0 & 2 \end{bmatrix} \begin{bmatrix} 1/\sqrt{2} & 1/\sqrt{2} \\ 1/\sqrt{2} & -1/\sqrt{2} \end{bmatrix} \right)$$

$\lambda = 10 \pm 6i$ , so we find the square roots by DeMoivre's formula ( $\sqrt{r}(\cos \theta/2 + i \sin \theta/2)$ )

$$\begin{aligned}r &= 10^2 + 6^2 = 136 \\ \theta &= \arctan(6/10) \approx .27\end{aligned}$$

$$\begin{aligned}A &= \begin{bmatrix} 10 & -6 \\ 6 & 10 \end{bmatrix} \\ A &= \frac{1}{2} \begin{bmatrix} 1 & 1 \\ i & -i \end{bmatrix} \begin{bmatrix} \sqrt{136}(\cos .27 + i \sin .27) & 0 \\ 0 & \sqrt{136}(\cos .27 - i \sin .27) \end{bmatrix} \begin{bmatrix} 1 & i \\ 1 & -i \end{bmatrix}\end{aligned}$$

8.

If  $A$  is positive definite, it may be expressed as  $A = R^T R$ . then

$$\begin{aligned}B &= C^T A C \\ &= C^T R^T R C\end{aligned}$$

$R^T R$  is positive definite, as are all  $A^T A$  of full rank. given that  $C$  is invertible,  $C^T C$  is positive definite as well, so the product of the two matrices must be positive.

## 9.

The generalized Schwarz inequality. I think I have the steps backward, but the logic is sound.

$$\begin{aligned}
|x^T Ay|^2 &\leq (x^T Ax)(y^T Ay) \\
&\leq (x^T R^T R x)(y^T R^T R y) \\
&\leq (Rx)^2 (Ry)^2 \\
|(RxRy)^2| &\leq (Rx)^2 (Ry)^2 \\
|(x^T R^T R y)^2| &\leq (x^T R^T R x)(y^T R^T R y) \\
|x^T Ay|^2 &\leq x^T A x y^T A y
\end{aligned}$$

The right side is invariably positive, but the left only is for all vectors if  $A$  is positive definite or  $x = y$

## 10.

Values obviously 1 and 4, vectors the standard vectors.

The major axis cuts through the x-axis, the minor axis the y-axis.

## 11.

From the polynomial  $3u^2 - 2\sqrt{uv} + 2v^2$ , we have:

$$S = \begin{bmatrix} \sqrt{2}/\sqrt{3} & 1/\sqrt{3} \\ 1/\sqrt{3} & -\sqrt{2}/\sqrt{3} \end{bmatrix}$$

implying the eigenvector factorization:

$$4 \left( \frac{\sqrt{2}}{\sqrt{3}} u + \frac{v}{\sqrt{3}} \right)^2 + \left( \frac{u}{\sqrt{3}} - \frac{\sqrt{2}}{\sqrt{3}} v \right)^2$$

This factorization just uses the eigenvalues as factors of squares to stand in for the effect of transforming  $x^T x$  by  $A$ .

## 13.

Tests for a *negative* definite matrix:

I.  $x^T k x < 0$  for all real  $x$  II. All  $\lambda < 0$  III. All upper-left submatrices have determinants of alternating signs, starting with negative, and the overall determinant is negative. The overall determinant must be negative because the opposite-sign of the matrix is positive definite. IV. All pivots  $< 0$

## 14.

$A$ : positive, since pivots are 1, 1, 17.

$B$ : Indefinite, since the pivots are 1, 2, 1, -7, and the subdeterminants don't go into the right pattern.

$C$ : also indefinite.

$D$  positive, as inverse of a positive.

**15.**

- a. False.
- b. True. This matrix is similar to  $A$ , and similar matrices have the same eigenvalues.
- c. True, by similarity.
- d. True; this is just the diagonal matrix of  $E$  raised to each eigenvalue, always positive.

**17.**

The product  $a_{11}a_{22}\dots a_{nn}$  is the product of the trace of  $R^T R$ . When  $A$  is diagonal, then  $R = R^T = \sqrt{A}$ . By the volume interpretation, all of  $A$ 's columns are orthogonal, so the parallelepiped is a perfect hyper-rectangle.

$$\begin{aligned} R &= R^T \\ \det A &= \det(R^T R) \\ &= \det(R)^2 \end{aligned}$$

**19.**

Matrix 1 is

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

Matrix 2 fills in the remaining zeroes and is not positive definite because the third row now requires elimination to get the pivots..

**20.**

$$A = \begin{bmatrix} 2 & 2 & 0 \\ 2 & 5 & 3 \\ 0 & 3 & 8 \end{bmatrix}$$

$$\det(2) = 2$$

$$2(5) - 2(2) = 6$$

$$2(5(8) - 3(3)) - 2(2(8) - 3(0)) + 0(2(3) - 5(0)) = 2(40 - 9) - 2(16 - 0) = 30$$

Indeed, pivot 2 is  $6/2 = 3$  and pivot 3 is  $30/6 = 5$ .

**21.**

The quadratic form  $2(x_1^2 + x_1x_2 + x_3x_1 + 2x_2x_3 + 5x_2^2)$  is -71 if, for example,  $(x_1, x_2, x_3) = 1, -10, 1$

**22.**

If entry  $A_{jj}$  of a PD matrix is smaller than any  $\lambda$ , then  $A - a_{jj}I$  would be PD as well. Yet this matrix has a zero in the place of  $a_{jj}$ , so it cannot be PD.

**23.**

- a. It cannot have zero eigenvalues, always present in invertible matrices.
- b. All projection matrices have rank  $m$ , where  $m$  is the dimension of the subspace of projection. The only such matrix for which  $m = n$  is the projection onto  $R^n$  - the identity.
- c. The eigenvalues are the diagonal and are all positive.

d.Example:

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

**24.**

a.

$$\begin{aligned} \det A &= s(s^2 + 16) + 4(-4s + 16) - 4(16 + 4s) \\ &= s^3 - 16s \\ &= s(s + 4)(s - 4) \end{aligned}$$

$$s \geq 4$$

b.

$$\begin{aligned} \det A &= t(t^2 - 9) - 3(3t - 0) \\ &= t^3 - 25t \\ t^2 &> 5 \\ t &\neq 0, -5, 5 \quad t > -5 \end{aligned}$$

**25.**

The coefficients of the completion of the square for the ellipse equation are  $\frac{1}{\sqrt{\lambda}}$ . For  $9x^2 + 16y^2 = 1$  they are  $1/3$  and  $1/4$ .

This works because the ellipse formula can be written as a sum of  $x_i^T \lambda_x x_i$  for each eigenvector, computing the elements of the final dot product separately.

**26.**

The matrix is just

$$\begin{bmatrix} 1 & 1/2 \\ 1/2 & 1 \end{bmatrix}$$

So  $\lambda = (3/2, 1/2)$ , so the half axes are  $\sqrt{2}/\sqrt{3}, \sqrt{2}$ .

**27.**

```

cholesky <- function(C) {
  L <- diag(nrow = nrow(C))
  L[lower.tri(L)] <- C[lower.tri(C)]
  D <- diag(C)
  L %*% sqrt(D) %*% sqrt(D) %*% t(L)
}
C <- square(3, 1, 0, 2)
mat2latex(cholesky(C))

```

$$\begin{bmatrix} 3 & 5.44948974278318 \\ 5.44948974278318 & 9.89897948556636 \end{bmatrix}$$

```

A <- square(4, 8, 8, 25)
C <- square(1, 2, 0, 1) %*% sqrt(diag(x = c(4, 25)))
mat2latex(C)

```

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

28.

$$A = \begin{bmatrix} 9 & 0 & 0 \\ 0 & 1 & 2 \\ 0 & 2 & 8 \end{bmatrix}$$

$$C = L\sqrt{D} = \begin{bmatrix} 3 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 2 & 2\sqrt{2} \end{bmatrix}$$

$$A = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 2 & 2 \\ 1 & 2 & 7 \end{bmatrix}$$

$$L\sqrt{D} = C = \begin{bmatrix} 1 & 0 & 0 \\ 1 & 1 & 0 \\ 1 & 1 & \sqrt{5} \end{bmatrix}$$

29.

Writing out  $x^T Ax = x^T LDL^T x$  reveals the square completion -  $(c - b^2/a)^2$  is the second term. The formula squares the sum of  $x$  and the ratio of base  $b$  to  $a$ , then adds the square of the difference of the third base and  $b^2$  in units of  $a$ .

$$2(x + 2y)^2 + (10 - 2)y^2$$

$$2x^2 + 8y^2 + 8xy + 8y^2$$

$$2x^2 + 8xy + 16y^2$$

**30.**

- a.  $\det A = 2 \times 5 = 10$
- b.  $\lambda = 2, 5$
- c.  $[1 \pm i]$
- d. It admits an  $LDL^T$  factorization, which can be expressed  $CC^T$

**31.**

$$B = (x_1 + x_x + x_3)^2$$

**32.**

- a. No; second subdeterminant is  $1^2 - 1^2 = 0$ .
- b. All eigenvalues are 0 or positive, so positive semidefinite.

**35.**

It does. After eliminating one row we have:

$$\begin{bmatrix} 2.5 & 3 & 0 \\ 3 & 5.9 & 7 \\ 0 & 7 & 7.5 \end{bmatrix}$$

Clearly  $7(7/5.9) > 7.5$ .

**36.**

- a. The squared eigenvectors magically disappear because they're orthonormal

$$\begin{aligned} z &= a_1x_1 + \cdots + a_px_p = v_1Cy_1 + \cdots + b_qCy_q \\ z &= S\alpha \\ z^TAz &= z^TAS\alpha \\ &= z^T\left(\sum_{i=1}^p \lambda_i a_i x_i\right) \\ &= \lambda_1 a_1^2 + \cdots + \lambda_p a_p^2 \geq 0 \end{aligned}$$

Since

$$\begin{aligned} C^T ACy &= \mu y \\ ACy &= (C^T)^{-1} \mu y \\ z &= b_1Cy_1 + \cdots + b_1Cy_1 \\ z^TAz &= z^TA(b_1Cy_1 + \cdots + b_1Cy_1) \\ &= (b_1y_1^TC^T + \cdots + b_qy^TC^T)(\mu_1b_1(C^T)^{-1}y + \cdots + \mu_qb_1(C^T)^{-1}y) \\ &= \mu_1b_1^2 + \cdots + \mu_1b_q^2 \end{aligned}$$

b. Because these expressions of  $z$  are equal, they fail to hold if any term of either side is nonzero, since  $\lambda_p a_p^2$  is always positive and  $\mu_q b_q^2$  always negative. So all  $a, b$  are zero. And if the eigenvectors are independent, then  $p + q \leq n$ .

c.

$$n - p + n - q \leq n$$

$$2n \leq n + p + q$$

$$n \leq p + q$$

This is compatible with  $p + q \leq n$  only if  $p + q = n$ .

**37.**

If  $C$  is nonsingular, then only the zero vector solves  $C^T x = 0$ . Then the kernel of  $C^T A C$  is that of  $A C$ . Let  $Cx = y$ . Then  $C^T A y$  has the kernel of  $A$  and therefore its rank.

**39.**

$C$  has to be square.

Orthogonal eigenvectors for a similarity transformation:

$$M = \begin{bmatrix} 1 & 0 \\ 0 & 2 \end{bmatrix} \quad x_i = \begin{bmatrix} \sqrt{3}-1 \\ 1 \end{bmatrix} \quad x_j = \begin{bmatrix} \sqrt{3}+1 \\ -1 \end{bmatrix}$$

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

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

$$(3 + -1) + 2$$

$$0$$

**41.**

$$\begin{bmatrix} 6 & -3 \\ -3 & 6 \end{bmatrix} x = \frac{\lambda}{18} \begin{bmatrix} 4 & 1 \\ 1 & 4 \end{bmatrix} x$$

$$R = \begin{bmatrix} 1 & 1/4 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix} = \begin{bmatrix} 2 & 1/2 \\ 0 & 2 \end{bmatrix}$$

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

$$C^T A C = \begin{bmatrix} 12 & 0 \\ -6 & 57/2 \end{bmatrix}$$

The values are 216 and 513, the vectors the standard vectors.