

Section 5.3 Problems

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1.

2.

3.

4.

No.

If A and B are orthogonal square matrices, which others also are?

5.

$3A$: No

6.

$-B$: Yes

7.

AB : Yes; orthogonal transformations preserve orthogonality.

8.

$A + B$: No.

9.

B^{-1} : Yes.

10.

A^{10} : Yes

$B^{-1}AB$: yes ,all are orthogonal, and orthogonal transformations preserve orthogonality.

11.

A^T : Yes, since $A^T A = I$, and orthogonal transformations preserve orthogonality.

We consider the symmetries of the same matrices, granted they are symmetric and B is invertible.

12.

3A: Yes

13.

$-b$: Yes

14.

Yes.

15.

AB : No. $AB_{ij} = [A_i \cdot B_j]$, not necessarily equal to $AB_{ji} = [A_j \cdot B_i]$.

16.

Yes. If $A_{ij} = A_{ji}$ and $B_{ij} = B_{ji}$ where $i \neq j$, then $A_{ij} + B_{ij} = A_{ji} + B_{ji} \implies (A + B)_{ji} = (A + B)_{ji}$

Yes.

$$\begin{aligned} A + B &= A^T + B^T \\ &= (A + B)^T \end{aligned}$$

17.

B^{-1} : yes, since $[B_i^{-1}] [B_j] = [B_j^{-1}] [B_i] = 0$

18.

A^{10} : apparently so. The triangular element is consistent in the dot products

```
A <- square(2, 4, 4, -1)
A %% 2
```

```
      [,1] [,2]
[1,]    20    4
[2,]     4   17
```

19.

$2I + 3A - 4A^2$: yes, we showed all these are symmetric.

20.

AB^2A : no, as AB is not necessarily symmetric.

For arbitrary square matrices, which are symmetric?

21.

$A^T A$: yes, since element ij and ji are equal.

22.

BB^T : yes, for the same reason

23.

$A - A^T$: no, if A is non-symmetric.

24.

$A^T B A$: No. Can be rearranged to $B A A^T$, not necessarily symmetric.

25

.... I think?

$$\begin{aligned} & A^T B^T B A \\ & (B A)^T B A \\ & ((B A)^T B A)^T = (B A)^T B A \end{aligned}$$

26.

$$\begin{aligned} & (B(A + A^T)B^T)^T = B(B(A + A^T))^T \\ & = B(A^T + A)B^T \\ & = B(A + A^T)B^T \end{aligned}$$

27.

$$\begin{aligned} & (A v) \cdot w = v \cdot (A^T w) \\ & (A v)^T w = v^T (A^T w) \\ & (v^T A^T) w = v^T (A^T w) \\ & v^T A^T w = v^T A^T w \end{aligned}$$

28.

$$\begin{aligned}(Ax) \cdot (Ay) &= x \cdot y \\ (Ax)^T Ay &= x^T y \\ x^T A^T Ay &= x^T y \\ A^T Ay &= y \\ A^T A &= I \\ A^T &= A^{-1}\end{aligned}$$

Of course, the transpose is the inverse only for orthogonal transformations

29.

$$\begin{aligned}\frac{v \cdot w}{||v|| ||w||} &= \frac{(Av) \cdot (Aw)}{||Av|| ||Aw||} \\ \frac{v \cdot w}{||v|| ||w||} &= \frac{v \cdot w}{||Av|| ||Aw||}\end{aligned}$$

Given that orthogonal transformations preserve lengths as well as dot products, we're done.

30.

If A is a transformation $R^m \rightarrow R^n$ that preserves length, then

$$\begin{aligned}\sqrt{(Av) \cdot (Av)} &= \sqrt{v \cdot v} \\ (Av) \cdot (Av) &= v \cdot v \\ v^T A^T A &= v^T v \\ A^T A &= I_m\end{aligned}$$

which means A^T is the left inverse of A . This implies A has full column rank, which means it must have partial row rank if $n \neq m$. That means $A^T A$ is invertible but not AA^T (a $n \times n$ of a matrix with partial row rank, since the transpose preserves rank, and products cannot increase rank).

31.

The rows of an orthogonal A must also be orthonormal because A^T is also orthogonal.

32.

- a. As shown above, $A^T A = I_m$ implies full column rank, which means AA^T cannot equal I_n .
- b. But if $A^T A = I_n$ for an $n \times n$, then it is orthogonal, so $A^T = A^{-1} \implies AA^T = I_n$

33.

By multiplying out, we see that orthogonal matrices (for which the inverse is the transpose) satisfy the equations $a^2 + b^2 = c^2 + d^2 = 1$ and $ac = -bd$. That suggests the basis $\begin{bmatrix} a & 1 - a^2 \\ 1 - a^2 & -a \end{bmatrix}$. Opposite-signed diagonals also work.

34.

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

35.

36.

$$\begin{bmatrix} a \\ b \\ c \end{bmatrix} = \sqrt{2} \begin{bmatrix} -3/8 \\ -3/8 \\ 3/2 \end{bmatrix}$$

38.

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

A^2 must be symmetric, since $A_{ij}^2 = A_{ji}^2$, both negative.

39.

If we have a line spanned by a unit vector, entry ij of the projection matrix is $u_i u_j / \|u\|^2$, with the squares on the diagonal.

40.

41.

The projection onto the unit line in R^n is given by a matrix consisting entirely of $1/n$, since each element of the unit vector is $1/\sqrt{n}$.

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

42.

We know by now projection matrices are idempotent: they rebalance the elements of a vector so it becomes part of the subspace, but a vector already in the subspace has exactly the correct ratios of elements already.

43.

Given a unit vector in R^3 , the matrix $A = 2uu^T - I_3$ describes the reflection $(2P - I)$, while the opposite sign $B = I_3 - 2uu^T$ is

44.

Given an $n \times m$ matrix, the dimension of the image and of the kernel of the transpose sum to n , because all vectors in R^n belong to one of those two spaces.

45.

$\dim(\ker(A)) = \dim(\ker(A^T))$ for matrices for which rank is exactly $n/2$ - only then is R^n partitioned equally.

46.

Trivial.

$$\begin{aligned} M &= QR \\ R &= Q^{-1}M \\ R &= Q^T M \end{aligned}$$

47.

For $A = QR$, then:

$$\begin{aligned} A^T A &= (QR)^T QR \\ &= R^T Q^T QR \\ &= R^T R \end{aligned}$$

so $A^T A = R^T R$. This makes sense - each column of R decomposes vectors of A into projections along unit vectors and the residual v^\perp , so $R^T R$ collects the intersection of column vectors in the same way as $A^T A$.

48.

We can also write:

$$\begin{aligned} A &= QR \\ A^T &= R^T Q^T \end{aligned}$$

since Q^T is also orthogonal.

49.

50.

a. Element 1, 1 of the matrix can only be 1. The nonzero entries a, b of the second column must satisfy:

$$\begin{aligned} a + 0b &= 0 \\ a^2 + b^2 &= 1 \end{aligned}$$

which only $a = 0, b = \pm 1$ satisfy. Proceeding column by column and restricting b to be positive, that leaves only the identity.

b.

$$A = Q_1 R_1 = Q_2 R_2$$

$$Q_2^{-1} Q_1 R_1 = R_2$$

$$Q_2^{-1} Q_1 = R_2 R_1^{-1}$$

$Q_2^{-1} Q_1 = I$ because the product of the triangular matrices must be orthogonal, but the only orthogonal triangular matrix possible here is the identity, so $Q_2^{-1} = Q_1^{-1}$, so $Q_2 = Q_1$,

and the same for the R s.

51.

a.

$$Q_1 = Q_2 S$$

$$Q^T Q_1 = (Q_2 S)^T Q_2 S$$

$$I = S^T Q_2^T Q_2 S$$

$$I = S^T S$$

so S must be orthogonal.

b.

$$M = Q_1 R_1 = Q_2 R_2$$

$$Q_1 = Q_2 R_2 R_1^{-1}$$

$R_2 R_1^{-1}$ must be orthogonal for the reasons given above, and again the only possible orthogonal triangular matrix is the identity. so $Q_1 = Q_2$ and $R_1 = R_2$. ## 52.

53.

54.

55.

$$\frac{n^2 - n}{2}$$

56.

57.

Yes. $L^{-1}(A^T) = A$ is the transpose from $R^{m \times n}$ to $R^{n \times m}$

58.

The kernel is 0. For the image, the diagonal remains the same and the off-diagonal element ij is $\frac{A_{ij}+A_{ji}}{2}$, so the resulting matrix is symmetric

59.

The kernel is all symmetric matrices. The image is a skew-symmetric matrix with a zero diagonal and element ij is $\frac{A_{ij}-A_{ji}}{2}$.

60.

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

61.

Applying the transformation zeroes out all but the third element of the basis, which gets doubled:

$$T\left(\begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix}\right) = \begin{bmatrix} 0 & 2 \\ -2 & 0 \end{bmatrix}$$

So the coordinates vector is $\begin{bmatrix} 0 \\ 0 \\ 0 \\ 2 \end{bmatrix}$ and the matrix $\begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 2 \end{bmatrix}$

62.

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64.

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72.

The projection onto

$$\begin{bmatrix} 1 \\ a \\ a^2 \\ \vdots \\ a^{n-1} \end{bmatrix}$$

is a Hankel matrix (positive sloping diagonals) of the same element) because the first column multiplies each element of the vector by 1, the second by a , and so on. This ensures that the diagonal elements are the same as the echelon diagonals.

73.