

# Extra Exercises

Nicholas Hoell

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These should challenge you a bit beyond what you've done in textbook exercises. A few of them, which are marked, are quite challenging.

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## 1 True/False

1. Let  $A$  be a matrix. The pivot columns of the reduced row echelon form of  $A$  are a basis for the row space of the transpose of  $A$ .
2. If  $A$  is invertible then  $\text{rank}(A) = \text{rank}(A^{-1})$ .
3. The rank of a matrix must be at least as big as the dimension of its nullspace.
4. A nonzero element of  $\text{Nul}(A)$ , for a square matrix  $A$ , must be an eigenvector of  $A$ .
5. If every entry of a square matrix  $A$  is positive then  $\det(A) \neq 0$ .
6. Let  $A$  be  $m \times k$  and  $B$  be  $k \times n$  matrices which have reduced row echelon matrices given by matrices  $M$  and  $N$  respectively. Then the reduced row echelon matrix of  $AB$  is given by the matrix  $MN$ .
7. Let  $A$  be invertible. Then, if  $\lambda$  is an eigenvalue of  $A$ ,  $\frac{1}{\lambda}$  is an eigenvalue of  $A^{-1}$ .
8. If  $A$  and  $B$  are  $2 \times 2$  then  $\det(2A + 3B) = 4 \det(A) + 9 \det(B)$ .
9. If  $A$  is  $n \times n$  and has eigenvalues  $\lambda_1, \dots, \lambda_k$  counted with multiplicities  $m_1, \dots, m_k$ , then  $\det(A) = \lambda_1^{m_1} \cdots \lambda_k^{m_k}$ .
10. If  $A$  is square and  $A^T = A^{-1}$  then  $\det(A) = 1$ .

11. Eigenvectors are linearly independent.
12. If  $A\mathbf{x} = \mathbf{0}$  has only the trivial solution then  $A$  is invertible.
13. If  $A$  is invertible, then so must be its reduced row echelon form.
14. The circle  $x^2 + y^2 = 1$  is the null space of some linear transformation  $T : \mathbb{R}^2 \rightarrow \mathbb{R}^2$ .
15. The line  $y = 6x$  is the null space of some linear transformation  $T : \mathbb{R}^2 \rightarrow \mathbb{R}^2$ .
16. The line  $y = -6x$  is the range of some linear transformation  $T : \mathbb{R}^2 \rightarrow \mathbb{R}^2$ .
17. If  $A$  is  $2 \times 2$  then  $\text{rank}(A) = \text{rank}(A^2)$ .
18. If  $\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3, \mathbf{v}_4$  are linearly independent vectors in subspace  $S$  then  $\dim(S) > 3$ .
19. Let  $A$  and  $B$  be  $n \times n$ . Then  $A$  and  $B$  are invertible if and only if  $AB$  is invertible.
20. If the characteristic polynomial of  $A$  is  $c_A(\lambda) = (\lambda - 4)^3(\lambda + 5)$  then  $A$  is  $4 \times 4$ .
21. If the characteristic polynomial of  $A$  is  $c_A(\lambda) = \lambda(\lambda - 1)^2(\lambda + 2)$  then there are no solutions to  $A\mathbf{x} = 3\mathbf{x}$ .
22. If 0 is an eigenvalue of  $A$  then  $A$  is not diagonalizable.
23.  $A$  is invertible if and only if  $\text{row}(A) = \text{col}(A)$ .
24. If  $A$  and  $B$  are the same size and each has rank 1 then  $\text{rank}(A + B) \geq 1$ .
25. If  $A$  and  $B$  are the same size and each have rank  $n$  then  $\text{rank}(A + B) = n$ .

## 2 Long Answer

1. (Some hard parts). This problem concerns planes and should challenge your intuitions.
  - A plane in  $\mathbb{R}^3$  is the solution set to an equation of the form  $ax + by + cz = d$ . Prove that two planes in  $\mathbb{R}^3$  have either zero or infinitely many intersection points.
  - A plane  $\Pi$  in  $\mathbb{R}^4$  is a set of the form  $\Pi = \{\mathbf{p} + s\mathbf{v}_1 + t\mathbf{v}_2 \mid s, t \in \mathbb{R}\}$ , where  $\mathbf{v}_1, \mathbf{v}_2$  are linearly independent. Prove that two planes in  $\mathbb{R}^4$  can meet at a single point.
  - Suppose two planes  $\Pi_1, \Pi_2$  in  $\mathbb{R}^4$  meet in a line  $L$ . Prove that there exists a translation<sup>1</sup> of one of the planes causing  $\Pi_1$  and  $\Pi_2$  to separate.

2. Let

$$A = \begin{pmatrix} 1 & 2 & 3 \\ 3 & 2 & 1 \end{pmatrix}$$

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<sup>1</sup>A translation of a plane  $\Pi = \{\mathbf{p} + s\mathbf{v}_1 + t\mathbf{v}_2 \mid s, t \in \mathbb{R}\}$  by the vector  $\mathbf{u}$  is the set  $\{\mathbf{u} + \mathbf{p} + s\mathbf{v}_1 + t\mathbf{v}_2 \mid s, t \in \mathbb{R}\}$ , often simply denoted  $\mathbf{u} + \Pi$ . It is a shift of all points in the plane by the vector  $\mathbf{u}$ .

- (a) Find a matrix  $B$  such that  $AB = I_2$  or show that such a  $B$  doesn't exist.
- (b) Find a matrix  $C$  such that  $CA = I_3$  or show that such a  $C$  doesn't exist.

3. Let

$$A = \begin{pmatrix} a & 1 & a & 0 & 0 & 0 \\ 0 & b & 1 & b & 0 & 0 \\ 0 & 0 & c & 1 & c & 0 \\ 0 & 0 & 0 & d & 1 & d \end{pmatrix}$$

where  $a, b, c, d$  are unspecified real numbers.

- (a) Prove that  $\text{rank}(A) > 2$ .
  - (b) Prove that if  $a = d = 0$  and  $bc = 1$  then  $\text{rank}(A) = 3$ .
4. Let  $V$  be a subspace of dimension  $n$  and let  $S = \{v_1, \dots, v_k\}$  be a subset of  $V$ . Answer True/False to the following:
- (a) If  $S$  is a basis for  $V$  then  $k = n$ .
  - (b) If  $S$  spans  $V$  then  $k \leq n$ .
  - (c) If  $S$  is linearly independent then  $k \leq n$ .
  - (d) If  $S$  is linearly independent and  $k = n$  then  $S$  spans  $V$ .
  - (e) If  $S$  spans  $V$  and  $k = n$  then  $S$  is a basis for  $V$ .

5. Find

$$\det \begin{pmatrix} 1 & 2 & 3 & 4 & 5 \\ 2 & 1 & 2 & 3 & 4 \\ 3 & 2 & 1 & 2 & 3 \\ 4 & 3 & 2 & 1 & 2 \\ 5 & 4 & 3 & 2 & 1 \end{pmatrix}$$

6. For this problem, we'll be considering a transformation. Let  $\mathbf{d}$  be a fixed, nonzero vector in  $\mathbb{R}^n$ . Define the transformation  $\mathbf{p}_{\mathbf{d}} : \mathbb{R}^n \rightarrow \mathbb{R}^n$  by  $\mathbf{p}_{\mathbf{d}}(\mathbf{x}) = \frac{\mathbf{x} \cdot \mathbf{d}}{\|\mathbf{d}\|^2} \mathbf{d}$
- (a) Prove that  $\mathbf{p}_{\mathbf{d}}$  is linear and find its standard matrix.
  - (b) Prove that  $\mathbf{p}_{\mathbf{d}} \circ \mathbf{p}_{\mathbf{d}} = \text{id}$
  - (c) Prove that  $\mathbf{x} - \mathbf{p}_{\mathbf{d}}(\mathbf{x})$  is orthogonal to  $\mathbf{d}$  for all  $\mathbf{x} \in \mathbb{R}^n$ .
  - (d) Use part (b) to find an example of a nonzero  $4 \times 4$  matrix, other than the identity, satisfying  $A^2 = A$ .
7. Are all matrices LU-factorizable? If not, provide an example of a matrix which is not factorable in this way.
8. A square matrix  $A$  is said to be *anti-symmetric* if  $A^T = -A$ .

- (a) Give two examples of  $2 \times 2$  anti-symmetric matrices.
- (b) Prove that every square matrix  $B$  can be written as a sum of a symmetric and an anti-symmetric matrix. Namely,  $B = S + A$  where  $S^T = S$ ,  $A^T = -A$ .
9. The *adjugate*  $\text{adj}(A)$  of a square matrix  $A$  is (uniquely) defined as the square matrix which satisfies  $\text{adj}(A)A = A\text{adj}(A) = \det(A)I$ . Prove that  $\text{adj}(A)^{-1} = \text{adj}(A^{-1})$  whenever  $A$  is invertible.
10. Let  $\mathbf{c} \neq \mathbf{0}$  be in  $\mathbb{R}^m$  and  $\mathbf{0} \neq \mathbf{r} \in \mathbb{R}^n$  and define  $A = \mathbf{c}\mathbf{r}^T$
- (a) Show that  $\text{col}(A) = \text{span}\{\mathbf{c}\}$  and  $\text{row}(A) = \text{span}\{\mathbf{r}\}$ .
- (b) Find  $\dim(\text{Nul}(A))$
- (c) Show that  $\text{Nul}(A) = \text{Nul}(\mathbf{r}^T)$
11. Let  $A$  be an  $n \times n$  matrix.
- (a) Show that  $A^2 = 0$  if and only if  $\text{col}(A) \subseteq \ker(A)$ .
- (b) Show that  $A^2 = 0$  implies  $\text{rank}(A) \leq \frac{n}{2}$
- (c) Find a matrix  $A$  such that  $\text{col}(A) = \text{null}(A)$
12. If the rows of a square matrix  $A$  all sum to the same number  $c$ , show that  $c$  is an eigenvalue of  $A$ .
13. If the columns of a square matrix  $A$  all sum to the same number  $c$ , show that  $c$  is an eigenvalue of  $A$ .
14. (Some hard parts). A square matrix is called *nilpotent* if there's an  $m \geq 1$  such that  $A^m = 0$ .
- (a) Find some examples of nonzero nilpotent matrices.
- (b) Prove that nilpotent matrices have only 0 as an eigenvalue.
- (c) Prove that  $c_A(x) = (-1)^n x^n$  for nilpotent matrices  $A$ .
- (d) If  $A, B$  are both nilpotent, of the same size, and  $AB = BA$  holds, then prove  $AB$  and  $A + B$  are both nilpotent as well.
- (e) Show that if  $A$  is nilpotent with  $A^{r+1} = 0$ , show that  $I - A$  is invertible and that its inverse is  $I + A + A^2 + \cdots + A^r$
15. Are all  $1 \times 1$  matrices diagonalizable?
16. Diagonalize  $\begin{bmatrix} 0 & 1 & 1 \\ 1 & 0 & 1 \\ 1 & 1 & 0 \end{bmatrix}$ .
17. Suppose  $A$  is diagonalizable.

- (a) Is  $A^n$  diagonalizable for  $n \geq 1$ ? If yes prove it, if no give a counterexample.
- (b) Is  $cA$  diagonalizable for all  $c \in \mathbb{R}$ ? If yes prove it, if no give a counterexample.
- (c) Is  $MAM^{-1}$  diagonalizable for any invertible  $M$  with same size as  $A$ ? If yes prove it, if no give a counterexample.
- (d) Is  $cI + A$  diagonalizable for all  $c \in \mathbb{R}$ ? If yes prove it, if no give a counterexample.
18. Are all elementary matrices diagonalizable? Why or why not.
19. Show that, for any  $n \times m$  matrix  $A$ , the matrix  $I_m + A^T A$  is invertible. Give two separate proofs.
20. True/False: Every square matrix can be written as the sum of two invertible matrices. If false give a counterexample. If true, prove it.
21. Let  $B$  be  $m \times n$  and  $AB$  be  $k \times n$ . Suppose that  $\text{rank}(AB) = \text{rank}(B)$ . Show that  $\ker B = \ker(AB)$ .
22. Let  $c_A(x)$  be the characteristic polynomial for matrix  $A$ .
- (a) Show  $c_{rA}(x) = r^n c_A(\frac{x}{r})$  for all  $r \neq 0$
- (b) Show that  $c_{A^2}(x^2) = c_A(x)c_A(-x)$
- (c) (*This may be quite tricky, but there may be an easy proof*) If  $A^m = 0$  for some  $m \geq 1$  show that  $c_A(x) = x^n$
23. Let  $A = [a_{ij}]$  be a square matrix with  $a_{ij} = 1$  for  $1 \leq i \leq n$  and  $1 \leq j \leq n$ . Is  $A$  diagonalizable? If yes, diagonalize it. If not, explain why not. *Hint: this can be done in a fast and easy way or a less fast and way less easy way. We suggest you choose the easy way.*
24. (Hard). Suppose that  $A$  is an  $n \times n$  matrix which satisfies  $A^2 - 3A + 2I = 0$ .
- (a) Show that the only possible eigenvalues of  $A$  are  $\lambda = 1, 2$ .
- (b) Show that  $A$  must be diagonalizable.
- (c) Find all possible matrices satisfying  $A^2 - 3A + 2I = 0$ .
25. (Very hard). Suppose that  $A$  is an  $m \times n$  matrix with  $n$  pivots. Show that there is an  $n \times n$  matrix  $Q$ , with real entries, satisfying

$$Q^2 = A^T A$$

Such a matrix  $Q$  would be called the *square root* of  $A^T A$ , namely  $Q = \sqrt{A^T A}$ .

26. (Some hard parts). Suppose that  $A$  is an  $m \times n$  matrix with  $n$  pivots.

- (a) Prove that there's a positive constant  $C$  such that  $\|A\mathbf{x}\| \leq C\|\mathbf{x}\|$  holds, for all  $\mathbf{x} \in \mathbb{R}^n$ .
- (b) Verify that  $\mathbf{x}^T A^T A \mathbf{x} = \|A\mathbf{x}\|^2$
- (c) Show that  $A^T A$  is diagonalizable with *strictly positive* eigenvalues. Thus,  $A^T A = P D P^{-1}$
- (d) Prove there's a positive constant  $c$  such that  $\|P\mathbf{y}\| > c\|\mathbf{y}\|$  holds for all  $\mathbf{y} \in \mathbb{R}^n$
- (e) Show that there's a positive constant  $d$  such that  $\|D\mathbf{z}\| > d\|\mathbf{z}\|$  holds for all  $\mathbf{z} \in \mathbb{R}^n$
- (f) Conclude that there's a positive constant  $\ell > 0$  such that  $\|A\mathbf{x}\| > \ell\|\mathbf{x}\|$  holds for all  $\mathbf{x} \in \mathbb{R}^n$