

## Answers to even numbered exercises

### 1.1

- 20.  $h = -2$ .
- 22.  $h = -5/3$ .
- 23. True, false, true, true.
- 24. True, false, false, false.
- 34.  $T_1 = 20, T_2 = 27.5, T_3 = 30, T_4 = 22.5$ .

### 1.2

- 20 (a)  $h = 9, k \neq 6$  (b)  $h \neq 9$ , any  $k$  (c)  $h = 9, k = 6$ .
- 21. False, false, true, true, false.
- 22. False, false, true, false, true.
- 24. No, since there will be a row of zeroes in the coefficient matrix to the left of this pivot.
- 26. Back substitution will produce a unique solution.
- 28. There should be a pivot in each column of the coefficient matrix, but not in the right-hand column of the augmented matrix.
- 30.  $x + y + z = 1$  and  $x + y + z = 2$ .
- 32. About half as  $n \rightarrow \infty$  (I think, but don't really care).
- 34. Matlab exercise.

### 1.3

- 23. True, false, true, true, false ( $\mathbf{u}$  and  $\mathbf{v}$  could be collinear).
- 24. True, true, false, true, true.

### 1.4

- 16. The echelon form of the augmented matrix is

$$\left[ \begin{array}{ccc|c} 1 & -3 & -4 & b_1 \\ 0 & -7 & -6 & b_2 + 3b_1 \\ 0 & 0 & 0 & b_1 + 2b_2 + b_3 \end{array} \right].$$

Thus there is a solution if and only if  $b_1 + 2b_2 + b_3 = 0$ .

- 23. False, true, false (true if we replace the words 'augmented matrix' by 'coefficient matrix'), true, true, true.
- 24. True, true, true, true, false, true.

### 1.5

- 23. True, false, false, false, false (since they don't say what  $\mathbf{p}$  is).
- 24. False, true, false, true, true (assuming some solution exists).
- 26. See 24(e).

### 1.7

- 21. True (assuming they mean ONLY the trivial solution), false, true, true.
- 22. True, false, true, false.
- 24. The second row must have all zeroes.
- 26. In other words  $\mathbf{a}_1, \mathbf{a}_2, \mathbf{a}_3$  are linearly independent. Then the echelon form has exactly one row of zeroes.
- 28. 5 (if there were a row of zeroes in the echelon form of the matrix, which we'll call  $A$ , then there would be no solution to  $A\mathbf{x} = \mathbf{b}$  for some  $\mathbf{b}$ ).
- 30.  $n$ .
- 34. True (really stupid question!).
- 36. False, whenever  $\mathbf{v}_1, \mathbf{v}_2$  and  $\mathbf{v}_4$  are already linearly dependent.
- 38. True. Any subset of a linearly independent set of vectors is linearly independent. Equivalently, any superset of a linearly dependent set of vectors is linearly dependent.

### 1.8

- 21. True, false, true, true, true.
- 22. True, true, false (it's an 'existence' question), true, true.

### 1.9

4.

$$\begin{bmatrix} \cos(-\pi/4) & -\sin(-\pi/4) \\ \sin(-\pi/4) & \cos(-\pi/4) \end{bmatrix} = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix}.$$

- 23. True, true, false, false, false.
- 24. False, true, true, false, true.
- 32.  $m$  (see Theorem 12).

## 2.1

15. False, false, true, true, false.

16. False (true without the + signs), True, False, False, True.

22. In general, the columns of an  $m \times n$  matrix  $M$  are linearly dependent if and only if there is a non-zero vector  $\mathbf{x} \in \mathbb{R}^n$  such that  $M\mathbf{x} = \mathbf{0}$ .

So suppose the columns of  $B$  are linearly dependent. Thus there exists a non-zero vector  $\mathbf{x}$  such that  $B\mathbf{x} = \mathbf{0}$ . Multiply both sides of this equation on the left by  $A$ , and we have  $A(B\mathbf{x}) = A \cdot \mathbf{0} = \mathbf{0}$ . But matrix multiplication is associative, so  $A(B\mathbf{x}) = (AB)\mathbf{x}$ . Thus  $(AB)\mathbf{x} = \mathbf{0}$  so, by the same reasoning as before, the columns of  $AB$  must be linearly dependent.

24. Let  $\mathbf{b}$  be given and multiply both sides of the equation  $AD = I_m$  on the right by  $\mathbf{b}$ . This yields  $(AD)\mathbf{b} = \mathbf{b}$ . By associativity of matrix multiplication, the left-hand side of this equals  $A(D\mathbf{b})$ . But then we have indeed a solution to  $A\mathbf{x} = \mathbf{b}$ , namely  $\mathbf{x} = D\mathbf{b}$ .

## 2.2

9. True (in the sense that a right-inverse must always be a left-inverse too, and vice versa), False, False (e.g.:  $A = \begin{bmatrix} 2 & 3 \\ 1 & 6 \end{bmatrix}$ ), True, True.

10. False, True, True, True, False.

12. Row reduction  $A \sim I$  corresponds to left-multiplication by  $A^{-1}$  (thought of as a product of elementary matrices). Performing the same row reduction on  $B$  thus results in  $A^{-1}B$ , v.s.v.

32. The matrix is not invertible, since the row operations  $R_2 \mapsto R_2 - 4R_1$ ,  $R_3 \mapsto R_3 + 2R_1$ ,  $R_3 \mapsto R_3 - 2R_2$  take it to the echelon form  $\begin{bmatrix} 1 & -2 & 1 \\ 0 & 1 & -1 \\ 0 & 0 & 0 \end{bmatrix}$ .

## 2.3

11. True, True, False (unless  $A$  is invertible), True, True.

12. True, True, True, False, True.

## 3.1

39. True, False (a subtle matter of terminology, since in the text the  $(i, j)$ -th cofactor is defined to be the number  $C_{i,j} = (-1)^{i+j} \det A_{ij}$ ).

40. False, False (true if we replace 'sum' by 'product').

### 3.2

**27.** True (since by ‘row replacement operation’ he means adding to a row some multiple of another : see page 197 and Theorem 3(a) on page 192), True, True, False.

**28.** True, False, False, False.

**32.**  $\det(rA) = r^n(\det A)$ .

### 3.3

**26.** A typical vector  $\mathbf{v}$  in the set  $\mathbf{p} + S$  is of the form  $\mathbf{v} = \mathbf{p} + \mathbf{s}$ , for some vector  $\mathbf{s} \in S$ . Applying  $T$  and using linearity we have  $T(\mathbf{v}) = T(\mathbf{p} + \mathbf{s}) = T(\mathbf{p}) + T(\mathbf{s})$ , which is just a typical element of  $T(\mathbf{p}) + T(S)$  since, by definition,  $T(S) = \{T(\mathbf{s}) : \mathbf{s} \in S\}$ .

**32.** Let  $T_1, T_2$  be the names of the tetrahedra with sides  $\mathbf{e}_1, \mathbf{e}_2, \mathbf{e}_3$  and  $\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3$  respectively. By the formula for the volume of a tetrahedron given in the text, we have that  $\text{Vol}(T_1) = 1/6$ , since it has perpendicular height one and its base is an equilateral triangle of side-length one, thus of area  $1/2$ .

Now the linear transformation defined by  $T(\mathbf{e}_i) = \mathbf{v}_i$ ,  $i = 1, 2, 3$  transforms  $T_1$  to  $T_2$ . By definition, the matrix of this transformation is  $M_T = [\mathbf{v}_1 \ \mathbf{v}_2 \ \mathbf{v}_3]$ , i.e.: the  $3 \times 3$  matrix whose columns are the  $\mathbf{v}$ -vectors. By the geometric definition of determinant, we have that  $\text{Vol } T_2 = |\det M_T| \cdot (\text{Vol } T_1)$ . Thus, by what we noted at the outset, it follows that

$$\text{Vol}(T_2) = \pm \frac{1}{6} \begin{vmatrix} | & | & | \\ \mathbf{v}_1 & \mathbf{v}_2 & \mathbf{v}_3 \\ | & | & | \end{vmatrix},$$

the sign depending on whether the determinant is positive or negative.

### 4.1

**4.** I will try so say this in words. Draw any line  $\mathcal{L}$  in the plane not passing through  $(0, 0)$ . Pick any point  $P$  on the line and let  $\mathbf{v}$  be the vector  $\vec{OP}$ . Consider  $2\mathbf{v}$ . This is the vector  $\vec{OQ}$ , where  $Q$  is the point along the line through  $O$  and  $P$ , which is twice as far away from  $O$  as is  $P$  and in the same direction. Clearly, this point is not on your line  $\mathcal{L}$ , thus proving that  $\mathcal{L}$  is not a subspace of  $\mathbb{R}^2$ .

**20 (a)** You need to know that a sum of two continuous functions is continuous, as is a scalar multiple of a continuous function (see Adams, Section 1.4, Theorem 6).

**(b)** Suppose  $f(a) = f(b)$  and  $g(a) = g(b)$ . Then, clearly,  $(f + g)(a) = (f + g)(b)$ . Also  $(cf)(a) = (cf)(b)$ , so the set of functions under consideration is closed under addition and scalar multiplication, hence a subspace of  $C[a, b]$ .

**23.** False, False (the opposite is true), False, True, False (they're digital, according to the waffle at the start of the chapter).

**24.** True, True, True (of itself),

False, though it is *isomorphic* to a subspace of  $\mathbb{R}^3$ , for example the subspace of all vectors whose  $z$ -component is zero,

False, since it doesn't say what  $\mathbf{u}$  and  $\mathbf{v}$  are. The statement would be true if it read instead : (ii) for any two vectors  $\mathbf{u}$  and  $\mathbf{v}$  in  $H$ , it is also the case that  $\mathbf{u} + \mathbf{v}$  is in  $H$  (iii) if  $\mathbf{u}$  is in  $H$  then so is  $c\mathbf{u}$ , for any scalar  $c$ .

**36.**  $\mathbf{y}$  is in  $\text{Col}(A)$  if and only if the system  $A\mathbf{x} = \mathbf{y}$  has a solution. So run the command  $A \setminus \mathbf{y}$  and see if you get an error message.

## 4.2

**25.** True, False, True, True (if he means that the equation is consistent for EVERY  $\mathbf{b}$ ), True, True.

**26.** True, True, False, True, True, True (don't bother yet as to why).

**30.** Let  $\mathbf{w}_1$  and  $\mathbf{w}_2$  be any two vectors in the range of  $T$  and  $c$  any scalar. We must show that both  $\mathbf{w}_1 + \mathbf{w}_2$  and  $c\mathbf{w}_1$  are in the range of  $T$ .

Since both  $\mathbf{w}_1$  and  $\mathbf{w}_2$  are in the range of  $T$  there exist, by definition, vectors  $\mathbf{v}_1$  and  $\mathbf{v}_2$  in  $V$  such that

$$T(\mathbf{v}_1) = \mathbf{w}_1, \quad T(\mathbf{v}_2) = \mathbf{w}_2.$$

But  $T$  is linear, thus

$$T(\mathbf{v}_1 + \mathbf{v}_2) = T(\mathbf{v}_1) + T(\mathbf{v}_2) = \mathbf{w}_1 + \mathbf{w}_2$$

and

$$T(c\mathbf{v}_1) = cT(\mathbf{v}_1) = c\mathbf{w}_1.$$

This equations show that both  $\mathbf{w}_1 + \mathbf{w}_2$  and  $c\mathbf{w}_1$  are in the range of  $T$ , as desired.

### 4.3

4. The matrix with these three vectors as its columns can be Gauss-reduced to the diagonal matrix

$$\begin{bmatrix} 2 & 1 & -7 \\ 0 & 1 & 1 \\ 0 & 0 & 12 \end{bmatrix}.$$

Thus, these three vectors are linearly independent and form a basis for  $\mathbb{R}^3$ .

10. The matrix can be Gauss-reduced to

$$\begin{bmatrix} 1 & 0 & -5 & 1 & 4 \\ 0 & 1 & -4 & 0 & 6 \\ 0 & 0 & 0 & 1 & -3 \end{bmatrix}.$$

Take  $x_3$  and  $x_5$  as free variables, and perform back substitution to get that a general element of the nullspace can be written as

$$x_3 \cdot \begin{bmatrix} 5 \\ 4 \\ 1 \\ 0 \\ 0 \end{bmatrix} + x_5 \cdot \begin{bmatrix} -7 \\ -6 \\ 0 \\ 3 \\ 1 \end{bmatrix}.$$

The two vectors above then form a basis for the nullspace.

21. False, False, True, False, True.

22. False, True, True, False, False (rather the corresponding columns in  $A$  itself).

30. Let  $A$  be the  $n \times k$  matrix which has these vectors as its columns. If these vectors formed a basis for  $\mathbb{R}^n$  then, in particular, they would be linearly independent. This would mean that  $\text{Nul}(A)$  would contain only the zero vector. But since  $A$  has more columns than rows, there will remain at least one free variable after Gauss elimination on  $A$ , and thus  $\text{Nul}(A)$  contains non-zero vectors (see Section 1.5, for example, though I don't know what theorem exactly he wants you to use (and it doesn't matter !)).

36, 38. Matlab exercises.

#### 4.4

10.  $\begin{bmatrix} 3 & 2 & 8 \\ -1 & 0 & -2 \\ 4 & -5 & 7 \end{bmatrix}.$

15. True, False, False ( $\mathbb{P}_3 \cong \mathbb{R}^4$ ).

16. True, False (other way round), True (namely, if the plane passes through the origin).

#### 4.5

6. Write out the subspace more explicitly as

$$\left\{ a \cdot \begin{bmatrix} 3 \\ 6 \\ -9 \\ -3 \end{bmatrix} + b \cdot \begin{bmatrix} 6 \\ -2 \\ 5 \\ 1 \end{bmatrix} + c \cdot \begin{bmatrix} -1 \\ -2 \\ 3 \\ 1 \end{bmatrix} \right\}.$$

Here the first vector is just -3 times the third one. So the space has dimension 2, and a basis is

$$\left\{ \begin{bmatrix} 6 \\ -2 \\ 5 \\ 1 \end{bmatrix}, \begin{bmatrix} -1 \\ -2 \\ 3 \\ 1 \end{bmatrix} \right\}.$$

14. Both are 3-dimensional.

19. True, False (unless it goes through the origin), False, False (unless  $S$  contains exactly  $n$  vectors), True.

20. False (see 4.1.24(d)), False (rather the number of FREE variables), False, False (see 19(d)), True.

29. True, True, True.

30. False, True, False.

#### 4.6

17. True, False (unless  $B$  was obtained from  $A$  without any row interchanges), True, False (unless  $A$  is square), ?? (don't know what he means and don't care!).

18. False (see 4.3.22(e)), False, True, True, True.

30. They must be equal, since consistency means that  $\mathbf{b}$  is a linear combination of the columns of  $A$ , hence adding  $\mathbf{b}$  as a column to the matrix does not increase the dimension of its column space.

#### 4.7

10.  $\begin{bmatrix} 2 & 3 \\ -5 & -8 \end{bmatrix}$  and  $\begin{bmatrix} 8 & 3 \\ -5 & -2 \end{bmatrix}$  respectively.  
11. False, True.  
12. True, False (rather it satisfies  $[\mathbf{x}]_{\mathcal{C}} = P[\mathbf{x}]_{\mathcal{B}}$ ).

#### 5.1

6. Compute

$$\begin{bmatrix} 3 & 6 & 7 \\ 3 & 3 & 7 \\ 5 & 6 & 5 \end{bmatrix} \begin{bmatrix} 1 \\ -2 \\ 1 \end{bmatrix} = \begin{bmatrix} -2 \\ 4 \\ -2 \end{bmatrix} = (-2) \cdot \begin{bmatrix} 1 \\ -2 \\ 1 \end{bmatrix}.$$

Thus, yes, it is an eigenvector, with corresponding eigenvalue  $-2$ .

21. False ( $\mathbf{x}$  must be non-zero), True, True, True, False (row operations can change the eigenvalues of a matrix).  
22. False (it's true if  $\mathbf{x}$  is not the zero vector), False (opposite true), True, False, True.

#### 5.2

18.  $h = 6$ .

20. We know that for any matrix  $M$  it holds that  $\det M = \det M^T$ . Let  $\lambda$  be a scalar. Then

$$\det(A - \lambda I) = \det(A - \lambda I)^T = \det(A^T - \lambda I).$$

Thus  $\det(A - \lambda I) = 0$  if and only if  $\det(A^T - \lambda I) = 0$ . In other words,  $\lambda$  is an eigenvalue of  $A$  if and only if it is an eigenvalue of  $A^T$ , v.s.v.

21. False, False, True, False (rather  $-5$  is then an eigenvalue).

22. False (the volume equals  $|\det A|$ ), False, True,

False : as an example, take  $\begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix}$ . This is diagonal, so its only eigenvalue

is  $\lambda = 1$ . The row replacement  $R_2 \mapsto R_2 - R_1$  produces the matrix  $\begin{bmatrix} 1 & 1 \\ -1 & 0 \end{bmatrix}$ .

One can check that the characteristic polynomial for this matrix is  $\lambda^2 - \lambda + 1$ , so there are two complex eigenvalues  $\lambda_{1,2} = \frac{1}{2}(1 \pm \sqrt{3}i)$ .

**24.** Similarity means that there exists an invertible matrix  $P$  such that  $B = P^{-1}AP$ . Then

$$\det B = \det(P^{-1}AP) = (\det P^{-1})(\det A)(\det P) = \left(\frac{1}{\det P}\right) (\det A)(\det P) = \det A, \quad \text{v.s.v.}$$

**30.** Blah ...

### 5.3

**21.** False ( $D$  must be diagonal), True, False, False.

**22.** False (true if we add the words 'linearly independent'), False (converse true), True, False.

### 5.7

**6.** The solution is

$$\mathbf{x}(t) = \begin{bmatrix} \mathbf{x}_1(t) \\ \mathbf{x}_2(t) \end{bmatrix} = \begin{bmatrix} 5e^{-t} - 2e^{-2t} \\ 5e^{-t} - 3e^{-2t} \end{bmatrix} = 5 \begin{bmatrix} 1 \\ 1 \end{bmatrix} e^{-t} - \begin{bmatrix} 2 \\ 3 \end{bmatrix} e^{-2t}.$$

The origin is an attractor and the direction of greatest attraction is along the line  $2y = 3x$ .

### 6.1

**6.**  $\frac{5}{49} \begin{bmatrix} 6 \\ -2 \\ 3 \end{bmatrix}$ .

**8.** 7.

**19.** True, true, true, false (rather  $\text{Row}(A)$  is orthogonal to  $\text{Nul}(A)$ ), true.

**20.** True, false (rather  $|c|$ ), true, true, true.

**24.** We have

$$\|\mathbf{u} \pm \mathbf{v}\|^2 = (\mathbf{u} \pm \mathbf{v}) \cdot (\mathbf{u} \pm \mathbf{v}) = \mathbf{u} \cdot \mathbf{u} + \mathbf{v} \cdot \mathbf{v} \pm 2(\mathbf{u} \cdot \mathbf{v}) = \|\mathbf{u}\|^2 + \|\mathbf{v}\|^2 \pm 2(\mathbf{u} \cdot \mathbf{v}).$$

When we add, the terms  $\pm 2(\mathbf{u} \cdot \mathbf{v})$  cancel and we're left with the right-hand side.

**26.**  $W$  is the nullspace of the  $1 \times 3$  matrix with the single row  $\mathbf{u}^T$ . So he's probably referring to some theorem that says that the nullspace of an  $m \times n$  matrix is a subspace of  $\mathbb{R}^n$ . Geometrically,  $W$  is a plane through the origin with normal vector  $\mathbf{u}$ . Its equation is  $5x - 6y + 7z = 0$ .

**34.** Blah ...

## 6.2

23. True, true, false, true, false (rather  $\|\mathbf{y} - \hat{\mathbf{y}}\|$ ).  
24. False, false (true if  $\|\mathbf{u}_i\| = 1$  for each  $i = 1, \dots, p$ ), true, true, true.

## 6.3

21. True, true, false, true, true.  
22. True, true, true, false (rather  $\text{proj}_W \mathbf{y}$ ), false (true when  $n = p$ ).

## 6.4

17. False (true if  $c \neq 0$ ), true, true.  
18. True, true, true.

## 6.5

17. True, true, false (rather  $\geq$ ), true, true.  
18. True, false, true, false (true if  $A^T A$  is invertible), ?? (don't understand what he means by 'reliable'), True (I guess).

## 6.6

4.  $y = \frac{1}{10}(43 - 7x)$ .  
10 (a) The model is  $A\mathbf{x} = \mathbf{b}$  where

$$A = \begin{bmatrix} e^{-.02t_1} & e^{-.07t_1} \\ e^{-.02t_2} & e^{-.07t_2} \\ e^{-.02t_3} & e^{-.07t_3} \\ e^{-.02t_4} & e^{-.07t_4} \\ e^{-.02t_5} & e^{-.07t_5} \end{bmatrix}, \quad \mathbf{x} = \begin{bmatrix} M_A \\ M_B \end{bmatrix}, \quad \mathbf{b} = \begin{bmatrix} y_1 \\ y_2 \\ y_3 \\ y_4 \\ y_5 \end{bmatrix}.$$

## 7.1

2. Not symmetric.  
4. Symmetric.  
6. Not symmetric.  
24. Check directly that  $A\mathbf{v}_1 = 10\mathbf{v}_1$  and  $A\mathbf{v}_2 = \mathbf{v}_2$ . Thus we have at least two eigenvalues,  $\lambda_1 = 10$  and  $\lambda_2 = 1$ . The matrix  $A$  is symmetric, so we know it must be orthogonally diagonalisable. Therefore, there must be a

third eigenvector  $\mathbf{v}_3$ , which is orthogonal to both  $\mathbf{v}_1$  and  $\mathbf{v}_2$ . But, since we're working in  $\mathbb{R}^3$ , there is only one possibility for such a vector, up to a scalar multiple, namely we can take

$$\mathbf{v}_3 = \mathbf{v}_1 \times \mathbf{v}_2 = \begin{vmatrix} \vec{i} & \vec{j} & \vec{k} \\ -2 & 2 & 1 \\ 1 & 1 & 0 \end{vmatrix} = -\vec{i} + \vec{j} - 4\vec{k}.$$

Thus  $\mathbf{v}_3 = \begin{bmatrix} -1 \\ 1 \\ 4 \end{bmatrix}$  must be an eigenvector. Now check directly that  $A\mathbf{v}_3 = \mathbf{v}_3$ . Thus the eigenvalue here is also  $\lambda_2 = 1$ , so this eigenspace is two-dimensional.

We then have an orthogonal diagonalisation

$$A = PDP^T,$$

where

$$D = \begin{bmatrix} 10 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}, \quad P = \begin{bmatrix} | & | & | \\ \mathbf{u}_1 & \mathbf{u}_2 & \mathbf{u}_3 \\ | & | & | \end{bmatrix}$$

and  $\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3$  are the normalised eigenvectors, i.e.:

$$\mathbf{u}_1 = \frac{\mathbf{v}_1}{\|\mathbf{v}_1\|} = \frac{\mathbf{v}_1}{3}, \quad \mathbf{u}_2 = \frac{\mathbf{v}_2}{\|\mathbf{v}_2\|} = \frac{\mathbf{v}_2}{\sqrt{2}}, \quad \mathbf{u}_3 = \frac{\mathbf{v}_3}{\|\mathbf{v}_3\|} = \frac{\mathbf{v}_3}{\sqrt{18}}.$$

Thus finally

$$P = \begin{bmatrix} -2/3 & 1/\sqrt{2} & -1/\sqrt{18} \\ 2/3 & 1/\sqrt{2} & 1/\sqrt{18} \\ 1/3 & 0 & -4/\sqrt{18} \end{bmatrix}.$$

**26.** True, true, false, true.

**28.**

$$(\mathbf{A}\mathbf{x}) \cdot \mathbf{y} = (\mathbf{A}\mathbf{x})^T \mathbf{y} = (\mathbf{x}^T \mathbf{A}^T) \mathbf{y} \stackrel{\mathbf{A}=\mathbf{A}^T}{=} (\mathbf{x}^T \mathbf{A}) \mathbf{y} = \mathbf{x}^T (\mathbf{A}\mathbf{y}) = \mathbf{x} \cdot (\mathbf{A}\mathbf{y}), \quad \text{v.s.v.}$$