

# lyryx with Open Texts

# LINEAR ALGEBRA with Applications

## Open Edition



ADAPTABLE | ACCESSIBLE | AFFORDABLE

Adapted for

Emory University

Math 221

Linear Algebra

Sections 1 & 2

Lectured and adapted by

Le Chen

April 15, 2021

le.chen@emory.edu

Course page

[http://math.emory.edu/~lchen41/teaching/2021\\_Spring\\_Math221](http://math.emory.edu/~lchen41/teaching/2021_Spring_Math221)

by W. Keith Nicholson

Creative Commons License (CC BY-NC-SA)



# Contents

---

<b>1</b>	<b>Systems of Linear Equations</b>	<b>5</b>
1.1	Solutions and Elementary Operations . . . . .	6
1.2	Gaussian Elimination . . . . .	16
1.3	Homogeneous Equations . . . . .	28
	Supplementary Exercises for Chapter 1 . . . . .	37
<b>2</b>	<b>Matrix Algebra</b>	<b>39</b>
2.1	Matrix Addition, Scalar Multiplication, and Transposition . . . . .	40
2.2	Matrix-Vector Multiplication . . . . .	53
2.3	Matrix Multiplication . . . . .	72
2.4	Matrix Inverses . . . . .	91
2.5	Elementary Matrices . . . . .	109
2.6	Linear Transformations . . . . .	119
2.7	LU-Factorization . . . . .	135
<b>3</b>	<b>Determinants and Diagonalization</b>	<b>147</b>
3.1	The Cofactor Expansion . . . . .	148
3.2	Determinants and Matrix Inverses . . . . .	163
3.3	Diagonalization and Eigenvalues . . . . .	178
	Supplementary Exercises for Chapter 3 . . . . .	201
<b>4</b>	<b>Vector Geometry</b>	<b>203</b>
4.1	Vectors and Lines . . . . .	204
4.2	Projections and Planes . . . . .	223
4.3	More on the Cross Product . . . . .	244
4.4	Linear Operators on $\mathbb{R}^3$ . . . . .	251
	Supplementary Exercises for Chapter 4 . . . . .	260
<b>5</b>	<b>Vector Space <math>\mathbb{R}^n</math></b>	<b>263</b>
5.1	Subspaces and Spanning . . . . .	264
5.2	Independence and Dimension . . . . .	273
5.3	Orthogonality . . . . .	287
5.4	Rank of a Matrix . . . . .	297

5.5	Similarity and Diagonalization . . . . .	307
	Supplementary Exercises for Chapter 5 . . . . .	320
<b>6</b>	<b>Vector Spaces</b>	<b>321</b>
6.1	Examples and Basic Properties . . . . .	322
6.2	Subspaces and Spanning Sets . . . . .	333
6.3	Linear Independence and Dimension . . . . .	342
6.4	Finite Dimensional Spaces . . . . .	354
	Supplementary Exercises for Chapter 6 . . . . .	364
<b>7</b>	<b>Linear Transformations</b>	<b>365</b>
7.1	Examples and Elementary Properties . . . . .	366
7.2	Kernel and Image of a Linear Transformation . . . . .	374
7.3	Isomorphisms and Composition . . . . .	385
<b>8</b>	<b>Orthogonality</b>	<b>399</b>
8.1	Orthogonal Complements and Projections . . . . .	400
8.2	Orthogonal Diagonalization . . . . .	410
8.3	Positive Definite Matrices . . . . .	421
8.4	QR-Factorization . . . . .	427
8.5	Computing Eigenvalues . . . . .	431
8.6	The Singular Value Decomposition . . . . .	436
8.6.1	Singular Value Decompositions . . . . .	436
8.6.2	Fundamental Subspaces . . . . .	442
8.6.3	The Polar Decomposition of a Real Square Matrix . . . . .	445
8.6.4	The Pseudoinverse of a Matrix . . . . .	447

## 6.3 Linear Independence and Dimension

### Definition 6.4 Linear Independence and Dependence

As in  $\mathbb{R}^n$ , a set of vectors  $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$  in a vector space  $V$  is called **linearly independent** (or simply **independent**) if it satisfies the following condition:

$$\text{If } s_1\mathbf{v}_1 + s_2\mathbf{v}_2 + \cdots + s_n\mathbf{v}_n = \mathbf{0}, \quad \text{then } s_1 = s_2 = \cdots = s_n = 0.$$

A set of vectors that is not linearly independent is said to be **linearly dependent** (or simply **dependent**).

The **trivial linear combination** of the vectors  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$  is the one with every coefficient zero:

$$0\mathbf{v}_1 + 0\mathbf{v}_2 + \cdots + 0\mathbf{v}_n$$

This is obviously one way of expressing  $\mathbf{0}$  as a linear combination of the vectors  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$ , and they are linearly independent when it is the *only* way.

### Example 6.3.1

Show that  $\{1+x, 3x+x^2, 2+x-x^2\}$  is independent in  $\mathbf{P}_2$ .

**Solution.** Suppose a linear combination of these polynomials vanishes.

$$s_1(1+x) + s_2(3x+x^2) + s_3(2+x-x^2) = 0$$

Equating the coefficients of 1,  $x$ , and  $x^2$  gives a set of linear equations.

$$\begin{aligned} s_1 + \quad + 2s_3 &= 0 \\ s_1 + 3s_2 + s_3 &= 0 \\ s_2 - s_3 &= 0 \end{aligned}$$

The only solution is  $s_1 = s_2 = s_3 = 0$ .

### Example 6.3.2

Show that  $\{\sin x, \cos x\}$  is independent in the vector space  $\mathbf{F}[0, 2\pi]$  of functions defined on the interval  $[0, 2\pi]$ .

**Solution.** Suppose that a linear combination of these functions vanishes.

$$s_1(\sin x) + s_2(\cos x) = 0$$

This must hold for *all* values of  $x$  in  $[0, 2\pi]$  (by the definition of equality in  $\mathbf{F}[0, 2\pi]$ ).

Taking  $x = 0$  yields  $s_2 = 0$  (because  $\sin 0 = 0$  and  $\cos 0 = 1$ ). Similarly,  $s_1 = 0$  follows from taking  $x = \frac{\pi}{2}$  (because  $\sin \frac{\pi}{2} = 1$  and  $\cos \frac{\pi}{2} = 0$ ).

**Example 6.3.3**

Suppose that  $\{\mathbf{u}, \mathbf{v}\}$  is an independent set in a vector space  $V$ . Show that  $\{\mathbf{u} + 2\mathbf{v}, \mathbf{u} - 3\mathbf{v}\}$  is also independent.

**Solution.** Suppose a linear combination of  $\mathbf{u} + 2\mathbf{v}$  and  $\mathbf{u} - 3\mathbf{v}$  vanishes:

$$s(\mathbf{u} + 2\mathbf{v}) + t(\mathbf{u} - 3\mathbf{v}) = \mathbf{0}$$

We must deduce that  $s = t = 0$ . Collecting terms involving  $\mathbf{u}$  and  $\mathbf{v}$  gives

$$(s + t)\mathbf{u} + (2s - 3t)\mathbf{v} = \mathbf{0}$$

Because  $\{\mathbf{u}, \mathbf{v}\}$  is independent, this yields linear equations  $s + t = 0$  and  $2s - 3t = 0$ . The only solution is  $s = t = 0$ .

**Example 6.3.4**

Show that any set of polynomials of distinct degrees is independent.

**Solution.** Let  $p_1, p_2, \dots, p_m$  be polynomials where  $\deg(p_i) = d_i$ . By relabelling if necessary, we may assume that  $d_1 > d_2 > \dots > d_m$ . Suppose that a linear combination vanishes:

$$t_1 p_1 + t_2 p_2 + \dots + t_m p_m = 0$$

where each  $t_i$  is in  $\mathbb{R}$ . As  $\deg(p_1) = d_1$ , let  $ax^{d_1}$  be the term in  $p_1$  of highest degree, where  $a \neq 0$ . Since  $d_1 > d_2 > \dots > d_m$ , it follows that  $t_1 ax^{d_1}$  is the only term of degree  $d_1$  in the linear combination  $t_1 p_1 + t_2 p_2 + \dots + t_m p_m = 0$ . This means that  $t_1 ax^{d_1} = 0$ , whence  $t_1 a = 0$ , hence  $t_1 = 0$  (because  $a \neq 0$ ). But then  $t_2 p_2 + \dots + t_m p_m = 0$  so we can repeat the argument to show that  $t_2 = 0$ . Continuing, we obtain  $t_i = 0$  for each  $i$ , as desired.

**Example 6.3.5**

Suppose that  $A$  is an  $n \times n$  matrix such that  $A^k = 0$  but  $A^{k-1} \neq 0$ . Show that  $B = \{I, A, A^2, \dots, A^{k-1}\}$  is independent in  $\mathbf{M}_n$ .

**Solution.** Suppose  $r_0 I + r_1 A + r_2 A^2 + \dots + r_{k-1} A^{k-1} = 0$ . Multiply by  $A^{k-1}$ :

$$r_0 A^{k-1} + r_1 A^k + r_2 A^{k+1} + \dots + r_{k-1} A^{2k-2} = 0$$

Since  $A^k = 0$ , all the higher powers are zero, so this becomes  $r_0 A^{k-1} = 0$ . But  $A^{k-1} \neq 0$ , so  $r_0 = 0$ , and we have  $r_1 A + r_2 A^2 + \dots + r_{k-1} A^{k-1} = 0$ . Now multiply by  $A^{k-2}$  to conclude that  $r_1 = 0$ . Continuing, we obtain  $r_i = 0$  for each  $i$ , so  $B$  is independent.

The next example collects several useful properties of independence for reference.

**Example 6.3.6**

Let  $V$  denote a vector space.

1. If  $\mathbf{v} \neq \mathbf{0}$  in  $V$ , then  $\{\mathbf{v}\}$  is an independent set.
2. No independent set of vectors in  $V$  can contain the zero vector.

**Solution.**

1. Let  $t\mathbf{v} = \mathbf{0}$ ,  $t$  in  $\mathbb{R}$ . If  $t \neq 0$ , then  $\mathbf{v} = \frac{1}{t}t\mathbf{v} = \frac{1}{t}\mathbf{0} = \mathbf{0}$ , contrary to assumption. So  $t = 0$ .
2. If  $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k\}$  is independent and (say)  $\mathbf{v}_2 = \mathbf{0}$ , then  $0\mathbf{v}_1 + 1\mathbf{v}_2 + \dots + 0\mathbf{v}_k = \mathbf{0}$  is a nontrivial linear combination that vanishes, contrary to the independence of  $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k\}$ .

A set of vectors is independent if  $\mathbf{0}$  is a linear combination in a unique way. The following theorem shows that *every* linear combination of these vectors has uniquely determined coefficients, and so extends Theorem 5.2.1.

**Theorem 6.3.1**

Let  $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$  be a linearly independent set of vectors in a vector space  $V$ . If a vector  $\mathbf{v}$  has two (ostensibly different) representations

$$\begin{aligned}\mathbf{v} &= s_1\mathbf{v}_1 + s_2\mathbf{v}_2 + \dots + s_n\mathbf{v}_n \\ \mathbf{v} &= t_1\mathbf{v}_1 + t_2\mathbf{v}_2 + \dots + t_n\mathbf{v}_n\end{aligned}$$

as linear combinations of these vectors, then  $s_1 = t_1, s_2 = t_2, \dots, s_n = t_n$ . In other words, every vector in  $V$  can be written in a unique way as a linear combination of the  $\mathbf{v}_i$ .

**Proof.** Subtracting the equations given in the theorem gives

$$(s_1 - t_1)\mathbf{v}_1 + (s_2 - t_2)\mathbf{v}_2 + \dots + (s_n - t_n)\mathbf{v}_n = \mathbf{0}$$

The independence of  $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$  gives  $s_i - t_i = 0$  for each  $i$ , as required. □

The following theorem extends (and proves) Theorem 5.2.4, and is one of the most useful results in linear algebra.

**Theorem 6.3.2: Fundamental Theorem**

can be spanned by  $n$  vectors. If any set of  $m$  vectors in  $V$  is linearly independent, then  $m \leq n$ .

**Proof.** Let  $V = \text{span}\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$ , and suppose that  $\{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_m\}$  is an independent set in  $V$ . Then  $\mathbf{u}_1 = a_1\mathbf{v}_1 + a_2\mathbf{v}_2 + \dots + a_n\mathbf{v}_n$  where each  $a_i$  is in  $\mathbb{R}$ . As  $\mathbf{u}_1 \neq \mathbf{0}$  (Example 6.3.6), not all of the

$a_i$  are zero, say  $a_1 \neq 0$  (after relabelling the  $\mathbf{v}_i$ ). Then  $V = \text{span}\{\mathbf{u}_1, \mathbf{v}_2, \mathbf{v}_3, \dots, \mathbf{v}_n\}$  as the reader can verify. Hence, write  $\mathbf{u}_2 = b_1\mathbf{u}_1 + c_2\mathbf{v}_2 + c_3\mathbf{v}_3 + \dots + c_n\mathbf{v}_n$ . Then some  $c_i \neq 0$  because  $\{\mathbf{u}_1, \mathbf{u}_2\}$  is independent; so, as before,  $V = \text{span}\{\mathbf{u}_1, \mathbf{u}_2, \mathbf{v}_3, \dots, \mathbf{v}_n\}$ , again after possible relabelling of the  $\mathbf{v}_i$ . If  $m > n$ , this procedure continues until all the vectors  $\mathbf{v}_i$  are replaced by the vectors  $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n$ . In particular,  $V = \text{span}\{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n\}$ . But then  $\mathbf{u}_{n+1}$  is a linear combination of  $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n$  contrary to the independence of the  $\mathbf{u}_i$ . Hence, the assumption  $m > n$  cannot be valid, so  $m \leq n$  and the theorem is proved.  $\square$

If  $V = \text{span}\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$ , and if  $\{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_m\}$  is an independent set in  $V$ , the above proof shows not only that  $m \leq n$  but also that  $m$  of the (spanning) vectors  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$  can be replaced by the (independent) vectors  $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_m$  and the resulting set will still span  $V$ . In this form the result is called the **Steinitz Exchange Lemma**.



**Definition 6.5 Basis of a Vector Space**

As in  $\mathbb{R}^n$ , a set  $\{\mathbf{e}_1, \mathbf{e}_2, \dots, \mathbf{e}_n\}$  of vectors in a vector space  $V$  is called a **basis** of  $V$  if it satisfies the following two conditions:

1.  $\{\mathbf{e}_1, \mathbf{e}_2, \dots, \mathbf{e}_n\}$  is linearly independent
2.  $V = \text{span}\{\mathbf{e}_1, \mathbf{e}_2, \dots, \mathbf{e}_n\}$

Thus if a set of vectors  $\{\mathbf{e}_1, \mathbf{e}_2, \dots, \mathbf{e}_n\}$  is a basis, then *every* vector in  $V$  can be written as a linear combination of these vectors in a *unique* way (Theorem 6.3.1). But even more is true: Any two (finite) bases of  $V$  contain the same number of vectors.

**Theorem 6.3.3: Invariance Theorem**

Let  $\{\mathbf{e}_1, \mathbf{e}_2, \dots, \mathbf{e}_n\}$  and  $\{\mathbf{f}_1, \mathbf{f}_2, \dots, \mathbf{f}_m\}$  be two bases of a vector space  $V$ . Then  $n = m$ .

**Proof.** Because  $V = \text{span}\{\mathbf{e}_1, \mathbf{e}_2, \dots, \mathbf{e}_n\}$  and  $\{\mathbf{f}_1, \mathbf{f}_2, \dots, \mathbf{f}_m\}$  is independent, it follows from Theorem 6.3.2 that  $m \leq n$ . Similarly  $n \leq m$ , so  $n = m$ , as asserted.  $\square$

Theorem 6.3.3 guarantees that no matter which basis of  $V$  is chosen it contains the same number of vectors as any other basis. Hence there is no ambiguity about the following definition.

**Definition 6.6 Dimension of a Vector Space**

If  $\{\mathbf{e}_1, \mathbf{e}_2, \dots, \mathbf{e}_n\}$  is a basis of the nonzero vector space  $V$ , the number  $n$  of vectors in the basis is called the **dimension** of  $V$ , and we write

$$\dim V = n$$

The zero vector space  $\{\mathbf{0}\}$  is defined to have dimension 0:

$$\dim \{\mathbf{0}\} = 0$$

In our discussion to this point we have always assumed that a basis is nonempty and hence that the dimension of the space is at least 1. However, the zero space  $\{\mathbf{0}\}$  has *no* basis (by Example 6.3.6) so our insistence that  $\dim \{\mathbf{0}\} = 0$  amounts to saying that the *empty* set of vectors is a basis of  $\{\mathbf{0}\}$ . Thus the statement that “the dimension of a vector space is the number of vectors in any basis” holds even for the zero space.

We saw in Example 5.2.9 that  $\dim(\mathbb{R}^n) = n$  and, if  $\mathbf{e}_j$  denotes column  $j$  of  $I_n$ , that  $\{\mathbf{e}_1, \mathbf{e}_2, \dots, \mathbf{e}_n\}$  is a basis (called the standard basis). In Example 6.3.7 below, similar considerations apply to the space  $\mathbf{M}_{mn}$  of all  $m \times n$  matrices; the verifications are left to the reader.

**Example 6.3.7**

The space  $\mathbf{M}_{mn}$  has dimension  $mn$ , and one basis consists of all  $m \times n$  matrices with exactly one entry equal to 1 and all other entries equal to 0. We call this the **standard basis** of  $\mathbf{M}_{mn}$ .

**Example 6.3.8**

Show that  $\dim \mathbf{P}_n = n + 1$  and that  $\{1, x, x^2, \dots, x^n\}$  is a basis, called the **standard basis** of  $\mathbf{P}_n$ .

**Solution.** Each polynomial  $p(x) = a_0 + a_1x + \dots + a_nx^n$  in  $\mathbf{P}_n$  is clearly a linear combination of  $1, x, \dots, x^n$ , so  $\mathbf{P}_n = \text{span}\{1, x, \dots, x^n\}$ . However, if a linear combination of these vectors vanishes,  $a_0 + a_1x + \dots + a_nx^n = 0$ , then  $a_0 = a_1 = \dots = a_n = 0$  because  $x$  is an indeterminate. So  $\{1, x, \dots, x^n\}$  is linearly independent and hence is a basis containing  $n + 1$  vectors. Thus,  $\dim(\mathbf{P}_n) = n + 1$ .

**Example 6.3.9**

If  $\mathbf{v} \neq \mathbf{0}$  is any nonzero vector in a vector space  $V$ , show that  $\text{span}\{\mathbf{v}\} = \mathbb{R}\mathbf{v}$  has dimension 1.

**Solution.**  $\{\mathbf{v}\}$  clearly spans  $\mathbb{R}\mathbf{v}$ , and it is linearly independent by Example 6.3.6. Hence  $\{\mathbf{v}\}$  is a basis of  $\mathbb{R}\mathbf{v}$ , and so  $\dim \mathbb{R}\mathbf{v} = 1$ .

**Example 6.3.10**

Let  $A = \begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix}$  and consider the subspace

$$U = \{X \text{ in } \mathbf{M}_{22} \mid AX = XA\}$$

of  $\mathbf{M}_{22}$ . Show that  $\dim U = 2$  and find a basis of  $U$ .

**Solution.** It was shown in Example 6.2.3 that  $U$  is a subspace for any choice of the matrix  $A$ . In the present case, if  $X = \begin{bmatrix} x & y \\ z & w \end{bmatrix}$  is in  $U$ , the condition  $AX = XA$  gives  $z = 0$  and  $x = y + w$ . Hence each matrix  $X$  in  $U$  can be written

$$X = \begin{bmatrix} y+w & y \\ 0 & w \end{bmatrix} = y \begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix} + w \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

so  $U = \text{span } B$  where  $B = \left\{ \begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \right\}$ . Moreover, the set  $B$  is linearly independent (verify this), so it is a basis of  $U$  and  $\dim U = 2$ .

**Example 6.3.11**

Show that the set  $V$  of all symmetric  $2 \times 2$  matrices is a vector space, and find the dimension of  $V$ .

**Solution.** A matrix  $A$  is symmetric if  $A^T = A$ . If  $A$  and  $B$  lie in  $V$ , then

$$(A+B)^T = A^T + B^T = A+B \quad \text{and} \quad (kA)^T = kA^T = kA$$

using Theorem 2.1.2. Hence  $A+B$  and  $kA$  are also symmetric. As the  $2 \times 2$  zero matrix is also in  $V$ , this shows that  $V$  is a vector space (being a subspace of  $\mathbf{M}_{22}$ ). Now a matrix  $A$  is symmetric when entries directly across the main diagonal are equal, so each  $2 \times 2$  symmetric matrix has the form

$$\begin{bmatrix} a & c \\ c & b \end{bmatrix} = a \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} + b \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} + c \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$$

Hence the set  $B = \left\{ \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}, \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \right\}$  spans  $V$ , and the reader can verify that  $B$  is linearly independent. Thus  $B$  is a basis of  $V$ , so  $\dim V = 3$ .

It is frequently convenient to alter a basis by multiplying each basis vector by a nonzero scalar. The next example shows that this always produces another basis. The proof is left as Exercise 6.3.22.

**Example 6.3.12**

Let  $B = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$  be nonzero vectors in a vector space  $V$ . Given nonzero scalars  $a_1, a_2, \dots, a_n$ , write  $D = \{a_1\mathbf{v}_1, a_2\mathbf{v}_2, \dots, a_n\mathbf{v}_n\}$ . If  $B$  is independent or spans  $V$ , the same is true of  $D$ . In particular, if  $B$  is a basis of  $V$ , so also is  $D$ .

## Exercises for 6.3

---

**Exercise 6.3.1** Show that each of the following sets of vectors is independent.

a.  $\{1+x, 1-x, x+x^2\}$  in  $\mathbf{P}_2$

b.  $\{x^2, x+1, 1-x-x^2\}$  in  $\mathbf{P}_2$

c.  $\left\{ \begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 1 & 0 \\ 1 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 1 & -1 \end{bmatrix}, \begin{bmatrix} 0 & 1 \\ 0 & 1 \end{bmatrix} \right\}$   
in  $\mathbf{M}_{22}$

d.  $\left\{ \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix}, \begin{bmatrix} 1 & 0 \\ 1 & 1 \end{bmatrix}, \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} \right\}$   
in  $\mathbf{M}_{22}$

b. If  $ax^2 + b(x+1) + c(1-x-x^2) = 0$ , then  $a+c = 0$ ,  $b-c = 0$ ,  $b+c = 0$ , so  $a = b = c = 0$ .

d. If  $a \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix} + b \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix} + c \begin{bmatrix} 1 & 0 \\ 1 & 1 \end{bmatrix} + d \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$ , then  $a + c + d = 0$ ,  $a + b + d = 0$ ,  $a + b + c = 0$ , and  $b + c + d = 0$ , so  $a = b = c = d = 0$ .

**Exercise 6.3.2** Which of the following subsets of  $V$  are independent?

- a.  $V = \mathbf{P}_2$ ;  $\{x^2 + 1, x + 1, x\}$   
 b.  $V = \mathbf{P}_2$ ;  $\{x^2 - x + 3, 2x^2 + x + 5, x^2 + 5x + 1\}$   
 c.  $V = \mathbf{M}_{22}$ ;  $\left\{ \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix}, \begin{bmatrix} 1 & 0 \\ 1 & 1 \end{bmatrix}, \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \right\}$   
 d.  $V = \mathbf{M}_{22}$ ;  
 $\left\{ \begin{bmatrix} -1 & 0 \\ 0 & -1 \end{bmatrix}, \begin{bmatrix} 1 & -1 \\ -1 & 1 \end{bmatrix}, \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}, \begin{bmatrix} 0 & -1 \\ -1 & 0 \end{bmatrix} \right\}$   
 e.  $V = \mathbf{F}[1, 2]$ ;  $\{\frac{1}{x}, \frac{1}{x^2}, \frac{1}{x^3}\}$   
 f.  $V = \mathbf{F}[0, 1]$ ;  $\left\{ \frac{1}{x^2+x-6}, \frac{1}{x^2-5x+6}, \frac{1}{x^2-9} \right\}$

b.  $3(x^2 - x + 3) - 2(2x^2 + x + 5) + (x^2 + 5x + 1) = 0$

d.  $2 \begin{bmatrix} -1 & 0 \\ 0 & -1 \end{bmatrix} + \begin{bmatrix} 1 & -1 \\ -1 & 1 \end{bmatrix} + \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$

f.  $\frac{5}{x^2+x-6} + \frac{1}{x^2-5x+6} - \frac{6}{x^2-9} = 0$

**Exercise 6.3.3** Which of the following are independent in  $\mathbf{F}[0, 2\pi]$ ?

- a.  $\{\sin^2 x, \cos^2 x\}$   
 b.  $\{1, \sin^2 x, \cos^2 x\}$   
 c.  $\{x, \sin^2 x, \cos^2 x\}$

b. Dependent:  $1 - \sin^2 x - \cos^2 x = 0$

**Exercise 6.3.4** Find all values of  $a$  such that the following are independent in  $\mathbb{R}^3$ .

a.  $\{(1, -1, 0), (a, 1, 0), (0, 2, 3)\}$

b.  $\{(2, a, 1), (1, 0, 1), (0, 1, 3)\}$

b.  $x \neq -\frac{1}{3}$

**Exercise 6.3.5** Show that the following are bases of the space  $V$  indicated.

a.  $\{(1, 1, 0), (1, 0, 1), (0, 1, 1)\}$ ;  $V = \mathbb{R}^3$

b.  $\{(-1, 1, 1), (1, -1, 1), (1, 1, -1)\}$ ;  $V = \mathbb{R}^3$

c.  $\left\{ \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}, \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix}, \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} \right\}$ ;  
 $V = \mathbf{M}_{22}$

d.  $\{1 + x, x + x^2, x^2 + x^3, x^3\}$ ;  $V = \mathbf{P}_3$

b. If  $r(-1, 1, 1) + s(1, -1, 1) + t(1, 1, -1) = (0, 0, 0)$ , then  $-r + s + t = 0$ ,  $r - s + t = 0$ , and  $r - s - t = 0$ , and this implies that  $r = s = t = 0$ . This proves independence. To prove that they span  $\mathbb{R}^3$ , observe that  $(0, 0, 1) = \frac{1}{2}[(-1, 1, 1) + (1, -1, 1)]$  so  $(0, 0, 1)$  lies in  $\text{span}\{(-1, 1, 1), (1, -1, 1), (1, 1, -1)\}$ . The proof is similar for  $(0, 1, 0)$  and  $(1, 0, 0)$ .

d. If  $r(1+x) + s(x+x^2) + t(x^2+x^3) + ux^3 = 0$ , then  $r = 0$ ,  $r + s = 0$ ,  $s + t = 0$ , and  $t + u = 0$ , so  $r = s = t = u = 0$ . This proves independence. To show that they span  $\mathbf{P}_3$ , observe that  $x^2 = (x^2 + x^3) - x^3$ ,  $x = (x + x^2) - x^2$ , and  $1 = (1 + x) - x$ , so  $\{1, x, x^2, x^3\} \subseteq \text{span}\{1 + x, x + x^2, x^2 + x^3, x^3\}$ .

**Exercise 6.3.6** Exhibit a basis and calculate the dimension of each of the following subspaces of  $\mathbf{P}_2$ .

a.  $\{a(1+x) + b(x+x^2) \mid a \text{ and } b \text{ in } \mathbb{R}\}$

b.  $\{a + b(x+x^2) \mid a \text{ and } b \text{ in } \mathbb{R}\}$

c.  $\{p(x) \mid p(1) = 0\}$

d.  $\{p(x) \mid p(x) = p(-x)\}$

- b.  $\{1, x+x^2\}$ ; dimension = 2  
 d.  $\{1, x^2\}$ ; dimension = 2

**Exercise 6.3.7** Exhibit a basis and calculate the dimension of each of the following subspaces of  $\mathbf{M}_{22}$ .

- a.  $\{A \mid A^T = -A\}$   
 b.  $\left\{A \mid A \begin{bmatrix} 1 & 1 \\ -1 & 0 \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ -1 & 0 \end{bmatrix} A\right\}$   
 c.  $\left\{A \mid A \begin{bmatrix} 1 & 0 \\ -1 & 0 \end{bmatrix} = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}\right\}$   
 d.  $\left\{A \mid A \begin{bmatrix} 1 & 1 \\ -1 & 0 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ -1 & 1 \end{bmatrix} A\right\}$

- b.  $\left\{\begin{bmatrix} 1 & 1 \\ -1 & 0 \end{bmatrix}, \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}\right\}$ ; dimension = 2  
 d.  $\left\{\begin{bmatrix} 1 & 0 \\ 1 & 1 \end{bmatrix}, \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix}\right\}$ ; dimension = 2

**Exercise 6.3.8** Let  $A = \begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix}$  and define  $U = \{X \mid X \in \mathbf{M}_{22} \text{ and } AX = X\}$ .

- a. Find a basis of  $U$  containing  $A$ .  
 b. Find a basis of  $U$  not containing  $A$ .

- b.  $\left\{\begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}\right\}$

**Exercise 6.3.9** Show that the set  $\mathbb{C}$  of all complex numbers is a vector space with the usual operations, and find its dimension.

**Exercise 6.3.10**

- a. Let  $V$  denote the set of all  $2 \times 2$  matrices with equal column sums. Show that  $V$  is a subspace of  $\mathbf{M}_{22}$ , and compute  $\dim V$ .

- b. Repeat part (a) for  $3 \times 3$  matrices.  
 c. Repeat part (a) for  $n \times n$  matrices.

- b.  $\dim V = 7$

**Exercise 6.3.11**

- a. Let  $V = \{(x^2 + x + 1)p(x) \mid p(x) \text{ in } \mathbf{P}_2\}$ . Show that  $V$  is a subspace of  $\mathbf{P}_4$  and find  $\dim V$ . [Hint: If  $f(x)g(x) = 0$  in  $\mathbf{P}$ , then  $f(x) = 0$  or  $g(x) = 0$ .]  
 b. Repeat with  $V = \{(x^2 - x)p(x) \mid p(x) \text{ in } \mathbf{P}_3\}$ , a subset of  $\mathbf{P}_5$ .  
 c. Generalize.

- b.  $\{x^2 - x, x(x^2 - x), x^2(x^2 - x), x^3(x^2 - x)\}$ ;  $\dim V = 4$

**Exercise 6.3.12** In each case, either prove the assertion or give an example showing that it is false.

- a. Every set of four nonzero polynomials in  $\mathbf{P}_3$  is a basis.  
 b.  $\mathbf{P}_2$  has a basis of polynomials  $f(x)$  such that  $f(0) = 0$ .  
 c.  $\mathbf{P}_2$  has a basis of polynomials  $f(x)$  such that  $f(0) = 1$ .  
 d. Every basis of  $\mathbf{M}_{22}$  contains a noninvertible matrix.  
 e. No independent subset of  $\mathbf{M}_{22}$  contains a matrix  $A$  with  $A^2 = 0$ .  
 f. If  $\{\mathbf{u}, \mathbf{v}, \mathbf{w}\}$  is independent then,  $a\mathbf{u} + b\mathbf{v} + c\mathbf{w} = \mathbf{0}$  for some  $a, b, c$ .  
 g.  $\{\mathbf{u}, \mathbf{v}, \mathbf{w}\}$  is independent if  $a\mathbf{u} + b\mathbf{v} + c\mathbf{w} = \mathbf{0}$  for some  $a, b, c$ .  
 h. If  $\{\mathbf{u}, \mathbf{v}\}$  is independent, so is  $\{\mathbf{u}, \mathbf{u} + \mathbf{v}\}$ .  
 i. If  $\{\mathbf{u}, \mathbf{v}\}$  is independent, so is  $\{\mathbf{u}, \mathbf{v}, \mathbf{u} + \mathbf{v}\}$ .  
 j. If  $\{\mathbf{u}, \mathbf{v}, \mathbf{w}\}$  is independent, so is  $\{\mathbf{u}, \mathbf{v}\}$ .

- k. If  $\{\mathbf{u}, \mathbf{v}, \mathbf{w}\}$  is independent, so is  $\{\mathbf{u} + \mathbf{w}, \mathbf{v} + \mathbf{w}\}$ .
- l. If  $\{\mathbf{u}, \mathbf{v}, \mathbf{w}\}$  is independent, so is  $\{\mathbf{u} + \mathbf{v} + \mathbf{w}\}$ .
- m. If  $\mathbf{u} \neq \mathbf{0}$  and  $\mathbf{v} \neq \mathbf{0}$  then  $\{\mathbf{u}, \mathbf{v}\}$  is dependent if and only if one is a scalar multiple of the other.
- n. If  $\dim V = n$ , then no set of more than  $n$  vectors can be independent.
- o. If  $\dim V = n$ , then no set of fewer than  $n$  vectors can span  $V$ .

- 
- b. No. Any linear combination  $f$  of such polynomials has  $f(0) = 0$ .
- d. No.  $\left\{ \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix}, \begin{bmatrix} 1 & 0 \\ 1 & 1 \end{bmatrix}, \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix} \right\}$ ; consists of invertible matrices.
- f. Yes.  $0\mathbf{u} + 0\mathbf{v} + 0\mathbf{w} = \mathbf{0}$  for every set  $\{\mathbf{u}, \mathbf{v}, \mathbf{w}\}$ .
- h. Yes.  $s\mathbf{u} + t(\mathbf{u} + \mathbf{v}) = \mathbf{0}$  gives  $(s+t)\mathbf{u} + t\mathbf{v} = \mathbf{0}$ , whence  $s+t = 0 = t$ .
- j. Yes. If  $r\mathbf{u} + s\mathbf{v} = \mathbf{0}$ , then  $r\mathbf{u} + s\mathbf{v} + 0\mathbf{w} = \mathbf{0}$ , so  $r = 0 = s$ .
- l. Yes.  $\mathbf{u} + \mathbf{v} + \mathbf{w} \neq \mathbf{0}$  because  $\{\mathbf{u}, \mathbf{v}, \mathbf{w}\}$  is independent.
- n. Yes. If  $I$  is independent, then  $|I| \leq n$  by the fundamental theorem because any basis spans  $V$ .

**Exercise 6.3.13** Let  $A \neq 0$  and  $B \neq 0$  be  $n \times n$  matrices, and assume that  $A$  is symmetric and  $B$  is skew-symmetric (that is,  $B^T = -B$ ). Show that  $\{A, B\}$  is independent.

**Exercise 6.3.14** Show that every set of vectors containing a dependent set is again dependent.

**Exercise 6.3.15** Show that every nonempty subset of an independent set of vectors is again independent.

If a linear combination of the subset vanishes, it is a linear combination of the vectors in the larger set

(coefficients outside the subset are zero) so it is trivial.

**Exercise 6.3.16** Let  $f$  and  $g$  be functions on  $[a, b]$ , and assume that  $f(a) = 1 = g(b)$  and  $f(b) = 0 = g(a)$ . Show that  $\{f, g\}$  is independent in  $\mathbf{F}[a, b]$ .

**Exercise 6.3.17** Let  $\{A_1, A_2, \dots, A_k\}$  be independent in  $\mathbf{M}_{mn}$ , and suppose that  $U$  and  $V$  are invertible matrices of size  $m \times m$  and  $n \times n$ , respectively. Show that  $\{UA_1V, UA_2V, \dots, UA_kV\}$  is independent.

**Exercise 6.3.18** Show that  $\{\mathbf{v}, \mathbf{w}\}$  is independent if and only if neither  $\mathbf{v}$  nor  $\mathbf{w}$  is a scalar multiple of the other.

**Exercise 6.3.19** Assume that  $\{\mathbf{u}, \mathbf{v}\}$  is independent in a vector space  $V$ . Write  $\mathbf{u}' = a\mathbf{u} + b\mathbf{v}$  and  $\mathbf{v}' = c\mathbf{u} + d\mathbf{v}$ , where  $a, b, c$ , and  $d$  are numbers. Show that  $\{\mathbf{u}', \mathbf{v}'\}$  is independent if and only if the matrix  $\begin{bmatrix} a & c \\ b & d \end{bmatrix}$  is invertible. [*Hint*: Theorem 2.4.5.]

Because  $\{\mathbf{u}, \mathbf{v}\}$  is linearly independent,  $s\mathbf{u}' + t\mathbf{v}' = \mathbf{0}$  is equivalent to  $\begin{bmatrix} a & c \\ b & d \end{bmatrix} \begin{bmatrix} s \\ t \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$ . Now apply Theorem 2.4.5.

**Exercise 6.3.20** If  $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k\}$  is independent and  $\mathbf{w}$  is not in  $\text{span}\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k\}$ , show that:

- $\{\mathbf{w}, \mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k\}$  is independent.
- $\{\mathbf{v}_1 + \mathbf{w}, \mathbf{v}_2 + \mathbf{w}, \dots, \mathbf{v}_k + \mathbf{w}\}$  is independent.

**Exercise 6.3.21** If  $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k\}$  is independent, show that  $\{\mathbf{v}_1, \mathbf{v}_1 + \mathbf{v}_2, \dots, \mathbf{v}_1 + \mathbf{v}_2 + \dots + \mathbf{v}_k\}$  is also independent.

**Exercise 6.3.22** Prove Example 6.3.12.

**Exercise 6.3.23** Let  $\{\mathbf{u}, \mathbf{v}, \mathbf{w}, \mathbf{z}\}$  be independent. Which of the following are dependent?

- $\{\mathbf{u} - \mathbf{v}, \mathbf{v} - \mathbf{w}, \mathbf{w} - \mathbf{u}\}$
- $\{\mathbf{u} + \mathbf{v}, \mathbf{v} + \mathbf{w}, \mathbf{w} + \mathbf{u}\}$
- $\{\mathbf{u} - \mathbf{v}, \mathbf{v} - \mathbf{w}, \mathbf{w} - \mathbf{z}, \mathbf{z} - \mathbf{u}\}$
- $\{\mathbf{u} + \mathbf{v}, \mathbf{v} + \mathbf{w}, \mathbf{w} + \mathbf{z}, \mathbf{z} + \mathbf{u}\}$

- 
- Independent.

- d. Dependent. For example,  $(\mathbf{u} + \mathbf{v}) - (\mathbf{v} + \mathbf{w}) + (\mathbf{w} + \mathbf{z}) - (\mathbf{z} + \mathbf{u}) = \mathbf{0}$ .

**Exercise 6.3.24** Let  $U$  and  $W$  be subspaces of  $V$  with bases  $\{\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3\}$  and  $\{\mathbf{w}_1, \mathbf{w}_2\}$  respectively. If  $U$  and  $W$  have only the zero vector in common, show that  $\{\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3, \mathbf{w}_1, \mathbf{w}_2\}$  is independent.

**Exercise 6.3.25** Let  $\{p, q\}$  be independent polynomials. Show that  $\{p, q, pq\}$  is independent if and only if  $\deg p \geq 1$  and  $\deg q \geq 1$ .

**Exercise 6.3.26** If  $z$  is a complex number, show that  $\{z, z^2\}$  is independent if and only if  $z$  is not real.

If  $z$  is not real and  $az + bz^2 = 0$ , then  $a + bz = 0$  ( $z \neq 0$ ). Hence if  $b \neq 0$ , then  $z = -ab^{-1}$  is real. So  $b = 0$ , and so  $a = 0$ . Conversely, if  $z$  is real, say  $z = a$ , then  $(-a)z + 1z^2 = 0$ , contrary to the independence of  $\{z, z^2\}$ .

**Exercise 6.3.27** Let  $B = \{A_1, A_2, \dots, A_n\} \subseteq \mathbf{M}_{mn}$ , and write  $B' = \{A_1^T, A_2^T, \dots, A_n^T\} \subseteq \mathbf{M}_{nm}$ . Show that:

- $B$  is independent if and only if  $B'$  is independent.
- $B$  spans  $\mathbf{M}_{mn}$  if and only if  $B'$  spans  $\mathbf{M}_{nm}$ .

**Exercise 6.3.28** If  $V = \mathbf{F}[a, b]$  as in Example 6.1.7, show that the set of constant functions is a subspace of dimension 1 ( $f$  is **constant** if there is a number  $c$  such that  $f(x) = c$  for all  $x$ ).

**Exercise 6.3.29**

- If  $U$  is an invertible  $n \times n$  matrix and  $\{A_1, A_2, \dots, A_{mn}\}$  is a basis of  $\mathbf{M}_{mn}$ , show that  $\{A_1U, A_2U, \dots, A_{mn}U\}$  is also a basis.
- Show that part (a) fails if  $U$  is not invertible. [*Hint*: Theorem 2.4.5.]

- If  $U\mathbf{x} = \mathbf{0}$ ,  $\mathbf{x} \neq \mathbf{0}$  in  $\mathbb{R}^n$ , then  $R\mathbf{x} = \mathbf{0}$  where  $R \neq 0$  is row 1 of  $U$ . If  $B \in \mathbf{M}_{mn}$  has each row equal to  $R$ , then  $B\mathbf{x} \neq \mathbf{0}$ . But if  $B = \sum r_i A_i U$ , then  $B\mathbf{x} = \sum r_i A_i U\mathbf{x} = \mathbf{0}$ . So  $\{A_i U\}$  cannot span  $\mathbf{M}_{mn}$ .

**Exercise 6.3.30** Show that  $\{(a, b), (a_1, b_1)\}$  is a basis of  $\mathbb{R}^2$  if and only if  $\{a + bx, a_1 + b_1x\}$  is a basis of  $\mathbf{P}_1$ .

**Exercise 6.3.31** Find the dimension of the subspace  $\text{span}\{1, \sin^2 \theta, \cos 2\theta\}$  of  $\mathbf{F}[0, 2\pi]$ .

**Exercise 6.3.32** Show that  $\mathbf{F}[0, 1]$  is not finite dimensional.

**Exercise 6.3.33** If  $U$  and  $W$  are subspaces of  $V$ , define their intersection  $U \cap W$  as follows:  $U \cap W = \{\mathbf{v} \mid \mathbf{v} \text{ is in both } U \text{ and } W\}$

- Show that  $U \cap W$  is a subspace contained in  $U$  and  $W$ .
- Show that  $U \cap W = \{\mathbf{0}\}$  if and only if  $\{\mathbf{u}, \mathbf{w}\}$  is independent for any nonzero vectors  $\mathbf{u}$  in  $U$  and  $\mathbf{w}$  in  $W$ .
- If  $B$  and  $D$  are bases of  $U$  and  $W$ , and if  $U \cap W = \{\mathbf{0}\}$ , show that  $B \cup D = \{\mathbf{v} \mid \mathbf{v} \text{ is in } B \text{ or } D\}$  is independent.

- If  $U \cap W = \mathbf{0}$  and  $r\mathbf{u} + s\mathbf{w} = \mathbf{0}$ , then  $r\mathbf{u} = -s\mathbf{w}$  is in  $U \cap W$ , so  $r\mathbf{u} = \mathbf{0} = s\mathbf{w}$ . Hence  $r = 0 = s$  because  $\mathbf{u} \neq \mathbf{0} \neq \mathbf{w}$ . Conversely, if  $\mathbf{v} \neq \mathbf{0}$  lies in  $U \cap W$ , then  $1\mathbf{v} + (-1)\mathbf{v} = \mathbf{0}$ , contrary to hypothesis.

**Exercise 6.3.34** If  $U$  and  $W$  are vector spaces, let  $V = \{(\mathbf{u}, \mathbf{w}) \mid \mathbf{u} \text{ in } U \text{ and } \mathbf{w} \text{ in } W\}$ .

- Show that  $V$  is a vector space if  $(\mathbf{u}, \mathbf{w}) + (\mathbf{u}_1, \mathbf{w}_1) = (\mathbf{u} + \mathbf{u}_1, \mathbf{w} + \mathbf{w}_1)$  and  $a(\mathbf{u}, \mathbf{w}) = (a\mathbf{u}, a\mathbf{w})$ .
- If  $\dim U = m$  and  $\dim W = n$ , show that  $\dim V = m + n$ .
- If  $V_1, \dots, V_m$  are vector spaces, let

$$\begin{aligned} V &= V_1 \times \dots \times V_m \\ &= \{(\mathbf{v}_1, \dots, \mathbf{v}_m) \mid \mathbf{v}_i \in V_i \text{ for each } i\} \end{aligned}$$

denote the space of  $n$ -tuples from the  $V_i$  with componentwise operations (see Exercise 6.1.17). If  $\dim V_i = n_i$  for each  $i$ , show that  $\dim V = n_1 + \dots + n_m$ .

**Exercise 6.3.35** Let  $\mathbf{D}_n$  denote the set of all functions  $f$  from the set  $\{1, 2, \dots, n\}$  to  $\mathbb{R}$ .

- a. Show that  $\mathbf{D}_n$  is a vector space with pointwise addition and scalar multiplication.
- b. Show that  $\{S_1, S_2, \dots, S_n\}$  is a basis of  $\mathbf{D}_n$  where, for each  $k = 1, 2, \dots, n$ , the function  $S_k$  is defined by  $S_k(k) = 1$ , whereas  $S_k(j) = 0$  if  $j \neq k$ .

**Exercise 6.3.36** A polynomial  $p(x)$  is called **even** if  $p(-x) = p(x)$  and **odd** if  $p(-x) = -p(x)$ . Let  $E_n$  and  $O_n$  denote the sets of even and odd polynomials in  $\mathbf{P}_n$ .

- a. Show that  $E_n$  is a subspace of  $\mathbf{P}_n$  and find  $\dim E_n$ .
- b. Show that  $O_n$  is a subspace of  $\mathbf{P}_n$  and find  $\dim O_n$ .

- b.  $\dim O_n = \frac{n}{2}$  if  $n$  is even and  $\dim O_n = \frac{n+1}{2}$  if  $n$  is odd.

**Exercise 6.3.37** Let  $\{\mathbf{v}_1, \dots, \mathbf{v}_n\}$  be independent in a vector space  $V$ , and let  $A$  be an  $n \times n$  matrix. Define  $\mathbf{u}_1, \dots, \mathbf{u}_n$  by

$$\begin{bmatrix} \mathbf{u}_1 \\ \vdots \\ \mathbf{u}_n \end{bmatrix} = A \begin{bmatrix} \mathbf{v}_1 \\ \vdots \\ \mathbf{v}_n \end{bmatrix}$$

(See Exercise 6.1.18.) Show that  $\{\mathbf{u}_1, \dots, \mathbf{u}_n\}$  is independent if and only if  $A$  is invertible.



