

# lyryx with Open Texts

# LINEAR ALGEBRA with Applications

## Open Edition



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Adapted for

Emory University

Math 221

Linear Algebra

Sections 1 & 2

Lectured and adapted by

Le Chen

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le.chen@emory.edu

Course page

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by W. Keith Nicholson

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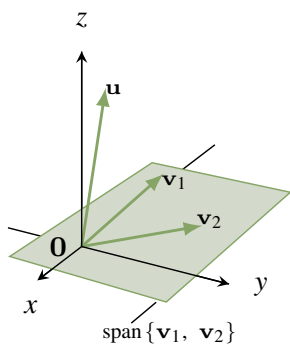
## 6.4 Finite Dimensional Spaces

Up to this point, we have had no guarantee that an arbitrary vector space *has* a basis—and hence no guarantee that one can speak *at all* of the dimension of  $V$ . However, Theorem 6.4.1 will show that any space that is spanned by a finite set of vectors has a (finite) basis: The proof requires the following basic lemma, of interest in itself, that gives a way to enlarge a given independent set of vectors.

### Lemma 6.4.1: Independent Lemma

Let  $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k\}$  be an independent set of vectors in a vector space  $V$ . If  $\mathbf{u} \in V$  but<sup>5</sup>  $\mathbf{u} \notin \text{span}\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k\}$ , then  $\{\mathbf{u}, \mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k\}$  is also independent.

**Proof.** Let  $t\mathbf{u} + t_1\mathbf{v}_1 + t_2\mathbf{v}_2 + \dots + t_k\mathbf{v}_k = \mathbf{0}$ ; we must show that all the coefficients are zero. First,  $t = 0$  because, otherwise,  $\mathbf{u} = -\frac{t_1}{t}\mathbf{v}_1 - \frac{t_2}{t}\mathbf{v}_2 - \dots - \frac{t_k}{t}\mathbf{v}_k$  is in  $\text{span}\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k\}$ , contrary to our assumption. Hence  $t = 0$ . But then  $t_1\mathbf{v}_1 + t_2\mathbf{v}_2 + \dots + t_k\mathbf{v}_k = \mathbf{0}$  so the rest of the  $t_i$  are zero by the independence of  $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k\}$ . This is what we wanted.  $\square$



Note that the converse of Lemma 6.4.1 is also true: if  $\{\mathbf{u}, \mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k\}$  is independent, then  $\mathbf{u}$  is not in  $\text{span}\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k\}$ .

As an illustration, suppose that  $\{\mathbf{v}_1, \mathbf{v}_2\}$  is independent in  $\mathbb{R}^3$ . Then  $\mathbf{v}_1$  and  $\mathbf{v}_2$  are not parallel, so  $\text{span}\{\mathbf{v}_1, \mathbf{v}_2\}$  is a plane through the origin (shaded in the diagram). By Lemma 6.4.1,  $\mathbf{u}$  is not in this plane if and only if  $\{\mathbf{u}, \mathbf{v}_1, \mathbf{v}_2\}$  is independent.

### Definition 6.7 Finite Dimensional and Infinite Dimensional Vector Spaces

A vector space  $V$  is called **finite dimensional** if it is spanned by a finite set of vectors. Otherwise,  $V$  is called **infinite dimensional**.

Thus the zero vector space  $\{\mathbf{0}\}$  is finite dimensional because  $\{\mathbf{0}\}$  is a spanning set.

### Lemma 6.4.2

Let  $V$  be a finite dimensional vector space. If  $U$  is any subspace of  $V$ , then any independent subset of  $U$  can be enlarged to a finite basis of  $U$ .

**Proof.**  $\text{span } I = U$  then  $I$  is already a basis of  $U$ . If  $\text{span } I \neq U$ , choose  $\mathbf{u}_1 \in U$  such that  $\mathbf{u}_1 \notin \text{span } I$ . Hence the set  $I \cup \{\mathbf{u}_1\}$  is independent by Lemma 6.4.1. If  $\text{span}(I \cup \{\mathbf{u}_1\}) = U$  we are done; otherwise choose  $\mathbf{u}_2 \in U$  such that  $\mathbf{u}_2 \notin \text{span}(I \cup \{\mathbf{u}_1\})$ . Hence  $I \cup \{\mathbf{u}_1, \mathbf{u}_2\}$  is independent, and the process

<sup>5</sup>If  $X$  is a set, we write  $a \in X$  to indicate that  $a$  is an element of the set  $X$ . If  $a$  is not an element of  $X$ , we write  $a \notin X$ .

continues. We claim that a basis of  $U$  will be reached eventually. Indeed, if no basis of  $U$  is ever reached, the process creates arbitrarily large independent sets in  $V$ . But this is impossible by the fundamental theorem because  $V$  is finite dimensional and so is spanned by a finite set of vectors.  $\square$

### Theorem 6.4.1

Let  $V$  be a finite dimensional vector space spanned by  $m$  vectors.

1.  $V$  has a finite basis, and  $\dim V \leq m$ .
2. Every independent set of vectors in  $V$  can be enlarged to a basis of  $V$  by adding vectors from any fixed basis of  $V$ .
3. If  $U$  is a subspace of  $V$ , then
  - a.  $U$  is finite dimensional and  $\dim U \leq \dim V$ .
  - b. If  $\dim U = \dim V$  then  $U = V$ .

### Proof.

1. If  $V = \{\mathbf{0}\}$ , then  $V$  has an empty basis and  $\dim V = 0 \leq m$ . Otherwise, let  $\mathbf{v} \neq \mathbf{0}$  be a vector in  $V$ . Then  $\{\mathbf{v}\}$  is independent, so (1) follows from Lemma 6.4.2 with  $U = V$ .
2. We refine the proof of Lemma 6.4.2. Fix a basis  $B$  of  $V$  and let  $I$  be an independent subset of  $V$ . If  $\text{span } I = V$  then  $I$  is already a basis of  $V$ . If  $\text{span } I \neq V$ , then  $B$  is not contained in  $I$  (because  $B$  spans  $V$ ). Hence choose  $\mathbf{b}_1 \in B$  such that  $\mathbf{b}_1 \notin \text{span } I$ . Hence the set  $I \cup \{\mathbf{b}_1\}$  is independent by Lemma 6.4.1. If  $\text{span}(I \cup \{\mathbf{b}_1\}) = V$  we are done; otherwise a similar argument shows that  $(I \cup \{\mathbf{b}_1, \mathbf{b}_2\})$  is independent for some  $\mathbf{b}_2 \in B$ . Continue this process. As in the proof of Lemma 6.4.2, a basis of  $V$  will be reached eventually.
3.
  - a. This is clear if  $U = \{\mathbf{0}\}$ . Otherwise, let  $\mathbf{u} \neq \mathbf{0}$  in  $U$ . Then  $\{\mathbf{u}\}$  can be enlarged to a finite basis  $B$  of  $U$  by Lemma 6.4.2, proving that  $U$  is finite dimensional. But  $B$  is independent in  $V$ , so  $\dim U \leq \dim V$  by the fundamental theorem.
  - b. This is clear if  $U = \{\mathbf{0}\}$  because  $V$  has a basis; otherwise, it follows from (2).  $\square$

Theorem 6.4.1 shows that a vector space  $V$  is finite dimensional if and only if it has a finite basis (possibly empty), and that every subspace of a finite dimensional space is again finite dimensional.

### Example 6.4.1

Enlarge the independent set  $D = \left\{ \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix}, \begin{bmatrix} 1 & 0 \\ 1 & 1 \end{bmatrix} \right\}$  to a basis of  $\mathbf{M}_{22}$ .

**Solution.** The standard basis of  $\mathbf{M}_{22}$  is  $\left\{ \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} \right\}$ , so including one of these in  $D$  will produce a basis by Theorem 6.4.1. In fact including *any* of these matrices in  $D$  produces an independent set (verify), and hence a basis by

Theorem 6.4.4. Of course these vectors are not the only possibilities, for example, including

$\begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix}$  works as well.

### Example 6.4.2

Find a basis of  $\mathbf{P}_3$  containing the independent set  $\{1+x, 1+x^2\}$ .

**Solution.** The standard basis of  $\mathbf{P}_3$  is  $\{1, x, x^2, x^3\}$ , so including two of these vectors will do. If we use 1 and  $x^3$ , the result is  $\{1, 1+x, 1+x^2, x^3\}$ . This is independent because the polynomials have distinct degrees (Example 6.3.4), and so is a basis by Theorem 6.4.1. Of course, including  $\{1, x\}$  or  $\{1, x^2\}$  would *not* work!

### Example 6.4.3

Show that the space  $\mathbf{P}$  of all polynomials is infinite dimensional.

**Solution.** For each  $n \geq 1$ ,  $\mathbf{P}$  has a subspace  $\mathbf{P}_n$  of dimension  $n+1$ . Suppose  $\mathbf{P}$  is finite dimensional, say  $\dim \mathbf{P} = m$ . Then  $\dim \mathbf{P}_n \leq \dim \mathbf{P}$  by Theorem 6.4.1, that is  $n+1 \leq m$ . This is impossible since  $n$  is arbitrary, so  $\mathbf{P}$  must be infinite dimensional.

The next example illustrates how (2) of Theorem 6.4.1 can be used.

### Example 6.4.4

If  $\mathbf{c}_1, \mathbf{c}_2, \dots, \mathbf{c}_k$  are independent columns in  $\mathbb{R}^n$ , show that they are the first  $k$  columns in some invertible  $n \times n$  matrix.

**Solution.** By Theorem 6.4.1, expand  $\{\mathbf{c}_1, \mathbf{c}_2, \dots, \mathbf{c}_k\}$  to a basis  $\{\mathbf{c}_1, \mathbf{c}_2, \dots, \mathbf{c}_k, \mathbf{c}_{k+1}, \dots, \mathbf{c}_n\}$  of  $\mathbb{R}^n$ . Then the matrix  $A = [\mathbf{c}_1 \ \mathbf{c}_2 \ \dots \ \mathbf{c}_k \ \mathbf{c}_{k+1} \ \dots \ \mathbf{c}_n]$  with this basis as its columns is an  $n \times n$  matrix and it is invertible by Theorem 5.2.3.

### Theorem 6.4.2

Let  $U$  and  $W$  be subspaces of the finite dimensional space  $V$ .

1. If  $U \subseteq W$ , then  $\dim U \leq \dim W$ .
2. If  $U \subseteq W$  and  $\dim U = \dim W$ , then  $U = W$ .

**Proof.** Since  $W$  is finite dimensional, (1) follows by taking  $V = W$  in part (3) of Theorem 6.4.1. Now assume  $\dim U = \dim W = n$ , and let  $B$  be a basis of  $U$ . Then  $B$  is an independent set in  $W$ . If  $U \neq W$ , then  $\text{span } B \neq W$ , so  $B$  can be extended to an independent set of  $n+1$  vectors in  $W$  by

Lemma 6.4.1. This contradicts the fundamental theorem (Theorem 6.3.2) because  $W$  is spanned by  $\dim W = n$  vectors. Hence  $U = W$ , proving (2).  $\square$

Theorem 6.4.2 is very useful. This was illustrated in Example 5.2.13 for  $\mathbb{R}^2$  and  $\mathbb{R}^3$ ; here is another example.

### Example 6.4.5

If  $a$  is a number, let  $W$  denote the subspace of all polynomials in  $\mathbf{P}_n$  that have  $a$  as a root:

$$W = \{p(x) \mid p(x) \in \mathbf{P}_n \text{ and } p(a) = 0\}$$

Show that  $\{(x-a), (x-a)^2, \dots, (x-a)^n\}$  is a basis of  $W$ .

**Solution.** Observe first that  $(x-a), (x-a)^2, \dots, (x-a)^n$  are members of  $W$ , and that they are independent because they have distinct degrees (Example 6.3.4). Write

$$U = \text{span} \{(x-a), (x-a)^2, \dots, (x-a)^n\}$$

Then we have  $U \subseteq W \subseteq \mathbf{P}_n$ ,  $\dim U = n$ , and  $\dim \mathbf{P}_n = n+1$ . Hence  $n \leq \dim W \leq n+1$  by Theorem 6.4.2. Since  $\dim W$  is an integer, we must have  $\dim W = n$  or  $\dim W = n+1$ . But then  $W = U$  or  $W = \mathbf{P}_n$ , again by Theorem 6.4.2. Because  $W \neq \mathbf{P}_n$ , it follows that  $W = U$ , as required.

A set of vectors is called **dependent** if it is *not* independent, that is if some nontrivial linear combination vanishes. The next result is a convenient test for dependence.

### Lemma 6.4.3: Dependent Lemma

A set  $D = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k\}$  of vectors in a vector space  $V$  is dependent if and only if some vector in  $D$  is a linear combination of the others.

**Proof.** Let  $\mathbf{v}_2$  (say) be a linear combination of the rest:  $\mathbf{v}_2 = s_1\mathbf{v}_1 + s_3\mathbf{v}_3 + \dots + s_k\mathbf{v}_k$ . Then

$$s_1\mathbf{v}_1 + (-1)\mathbf{v}_2 + s_3\mathbf{v}_3 + \dots + s_k\mathbf{v}_k = \mathbf{0}$$

is a nontrivial linear combination that vanishes, so  $D$  is dependent. Conversely, if  $D$  is dependent, let  $t_1\mathbf{v}_1 + t_2\mathbf{v}_2 + \dots + t_k\mathbf{v}_k = \mathbf{0}$  where some coefficient is nonzero. If (say)  $t_2 \neq 0$ , then  $\mathbf{v}_2 = -\frac{t_1}{t_2}\mathbf{v}_1 - \frac{t_3}{t_2}\mathbf{v}_3 - \dots - \frac{t_k}{t_2}\mathbf{v}_k$  is a linear combination of the others.  $\square$

Lemma 6.4.1 gives a way to enlarge independent sets to a basis; by contrast, Lemma 6.4.3 shows that spanning sets can be cut down to a basis.

### Theorem 6.4.3

Let  $V$  be a finite dimensional vector space. Any spanning set for  $V$  can be cut down (by deleting vectors) to a basis of  $V$ .



**Proof.** Since  $V$  is finite dimensional, it has a finite spanning set  $S$ . Among all spanning sets contained in  $S$ , choose  $S_0$  containing the smallest number of vectors. It suffices to show that  $S_0$  is independent (then  $S_0$  is a basis, proving the theorem). Suppose, on the contrary, that  $S_0$  is not independent. Then, by Lemma 6.4.3, some vector  $\mathbf{u} \in S_0$  is a linear combination of the set  $S_1 = S_0 \setminus \{\mathbf{u}\}$  of vectors in  $S_0$  other than  $\mathbf{u}$ . It follows that  $\text{span } S_0 = \text{span } S_1$ , that is,  $V = \text{span } S_1$ . But  $S_1$  has fewer elements than  $S_0$  so this contradicts the choice of  $S_0$ . Hence  $S_0$  is independent after all.  $\square$

Note that, with Theorem 6.4.1, Theorem 6.4.3 completes the promised proof of Theorem 5.2.6 for the case  $V = \mathbb{R}^n$ .

#### Example 6.4.6

Find a basis of  $\mathbf{P}_3$  in the spanning set  $S = \{1, x+x^2, 2x-3x^2, 1+3x-2x^2, x^3\}$ .

**Solution.** Since  $\dim \mathbf{P}_3 = 4$ , we must eliminate one polynomial from  $S$ . It cannot be  $x^3$  because the span of the rest of  $S$  is contained in  $\mathbf{P}_2$ . But eliminating  $1+3x-2x^2$  does leave a basis (verify). Note that  $1+3x-2x^2$  is the sum of the first three polynomials in  $S$ .

Theorems 6.4.1 and 6.4.3 have other useful consequences.

#### Theorem 6.4.4

*Let  $V$  be a vector space with  $\dim V = n$ , and suppose  $S$  is a set of exactly  $n$  vectors in  $V$ . Then  $S$  is independent if and only if  $S$  spans  $V$ .*

**Proof.** Assume first that  $S$  is independent. By Theorem 6.4.1,  $S$  is contained in a basis  $B$  of  $V$ . Hence  $|S| = n = |B|$  so, since  $S \subseteq B$ , it follows that  $S = B$ . In particular  $S$  spans  $V$ .

Conversely, assume that  $S$  spans  $V$ , so  $S$  contains a basis  $B$  by Theorem 6.4.3. Again  $|S| = n = |B|$  so, since  $S \supseteq B$ , it follows that  $S = B$ . Hence  $S$  is independent.  $\square$

One of independence or spanning is often easier to establish than the other when showing that a set of vectors is a basis. For example if  $V = \mathbb{R}^n$  it is easy to check whether a subset  $S$  of  $\mathbb{R}^n$  is orthogonal (hence independent) but checking spanning can be tedious. Here are three more examples.

#### Example 6.4.7

Consider the set  $S = \{p_0(x), p_1(x), \dots, p_n(x)\}$  of polynomials in  $\mathbf{P}_n$ . If  $\deg p_k(x) = k$  for each  $k$ , show that  $S$  is a basis of  $\mathbf{P}_n$ .

**Solution.** The set  $S$  is independent—the degrees are distinct—see Example 6.3.4. Hence  $S$  is a basis of  $\mathbf{P}_n$  by Theorem 6.4.4 because  $\dim \mathbf{P}_n = n + 1$ .

**Example 6.4.8**

Let  $V$  denote the space of all symmetric  $2 \times 2$  matrices. Find a basis of  $V$  consisting of invertible matrices.

**Solution.** We know that  $\dim V = 3$  (Example 6.3.11), so what is needed is a set of three invertible, symmetric matrices that (using Theorem 6.4.4) is either independent or spans  $V$ . The set  $\left\{ \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}, \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \right\}$  is independent (verify) and so is a basis of the required type.

**Example 6.4.9**

Let  $A$  be any  $n \times n$  matrix. Show that there exist  $n^2 + 1$  scalars  $a_0, a_1, a_2, \dots, a_{n^2}$  not all zero, such that

$$a_0I + a_1A + a_2A^2 + \cdots + a_{n^2}A^{n^2} = 0$$

where  $I$  denotes the  $n \times n$  identity matrix.

**Solution.** The space  $\mathbf{M}_n$  of all  $n \times n$  matrices has dimension  $n^2$  by Example 6.3.7. Hence the  $n^2 + 1$  matrices  $I, A, A^2, \dots, A^{n^2}$  cannot be independent by Theorem 6.4.4, so a nontrivial linear combination vanishes. This is the desired conclusion.

The result in Example 6.4.9 can be written as  $f(A) = 0$  where  $f(x) = a_0 + a_1x + a_2x^2 + \cdots + a_{n^2}x^{n^2}$ . In other words,  $A$  satisfies a nonzero polynomial  $f(x)$  of degree at most  $n^2$ . In fact we know that  $A$  satisfies a nonzero polynomial of degree  $n$  (this is the Cayley-Hamilton theorem—see Theorem ??), but the brevity of the solution in Example 6.4.6 is an indication of the power of these methods.

If  $U$  and  $W$  are subspaces of a vector space  $V$ , there are two related subspaces that are of interest, their **sum**  $U + W$  and their **intersection**  $U \cap W$ , defined by

$$\begin{aligned} U + W &= \{\mathbf{u} + \mathbf{w} \mid \mathbf{u} \in U \text{ and } \mathbf{w} \in W\} \\ U \cap W &= \{\mathbf{v} \in V \mid \mathbf{v} \in U \text{ and } \mathbf{v} \in W\} \end{aligned}$$

It is routine to verify that these are indeed subspaces of  $V$ , that  $U \cap W$  is contained in both  $U$  and  $W$ , and that  $U + W$  contains both  $U$  and  $W$ . We conclude this section with a useful fact about the dimensions of these spaces. The proof is a good illustration of how the theorems in this section are used.

**Theorem 6.4.5**

Suppose that  $U$  and  $W$  are finite dimensional subspaces of a vector space  $V$ . Then  $U + W$  is finite dimensional and

$$\dim(U + W) = \dim U + \dim W - \dim(U \cap W).$$

**Proof.** Since  $U \cap W \subseteq U$ , it has a finite basis, say  $\{\mathbf{x}_1, \dots, \mathbf{x}_d\}$ . Extend it to a basis  $\{\mathbf{x}_1, \dots, \mathbf{x}_d, \mathbf{u}_1, \dots, \mathbf{u}_m\}$  of  $U$  by Theorem 6.4.1. Similarly extend  $\{\mathbf{x}_1, \dots, \mathbf{x}_d\}$  to a basis  $\{\mathbf{x}_1, \dots, \mathbf{x}_d, \mathbf{w}_1, \dots, \mathbf{w}_p\}$  of  $W$ .

Then

$$U + W = \text{span}\{\mathbf{x}_1, \dots, \mathbf{x}_d, \mathbf{u}_1, \dots, \mathbf{u}_m, \mathbf{w}_1, \dots, \mathbf{w}_p\}$$

as the reader can verify, so  $U + W$  is finite dimensional. For the rest, it suffices to show that  $\{\mathbf{x}_1, \dots, \mathbf{x}_d, \mathbf{u}_1, \dots, \mathbf{u}_m, \mathbf{w}_1, \dots, \mathbf{w}_p\}$  is independent (verify). Suppose that

$$r_1\mathbf{x}_1 + \dots + r_d\mathbf{x}_d + s_1\mathbf{u}_1 + \dots + s_m\mathbf{u}_m + t_1\mathbf{w}_1 + \dots + t_p\mathbf{w}_p = \mathbf{0} \quad (6.1)$$

where the  $r_i$ ,  $s_j$ , and  $t_k$  are scalars. Then

$$r_1\mathbf{x}_1 + \dots + r_d\mathbf{x}_d + s_1\mathbf{u}_1 + \dots + s_m\mathbf{u}_m = -(t_1\mathbf{w}_1 + \dots + t_p\mathbf{w}_p)$$

is in  $U$  (left side) and also in  $W$  (right side), and so is in  $U \cap W$ . Hence  $(t_1\mathbf{w}_1 + \dots + t_p\mathbf{w}_p)$  is a linear combination of  $\{\mathbf{x}_1, \dots, \mathbf{x}_d\}$ , so  $t_1 = \dots = t_p = 0$ , because  $\{\mathbf{x}_1, \dots, \mathbf{x}_d, \mathbf{w}_1, \dots, \mathbf{w}_p\}$  is independent. Similarly,  $s_1 = \dots = s_m = 0$ , so (6.1) becomes  $r_1\mathbf{x}_1 + \dots + r_d\mathbf{x}_d = \mathbf{0}$ . It follows that  $r_1 = \dots = r_d = 0$ , as required.  $\square$

Theorem 6.4.5 is particularly interesting if  $U \cap W = \{\mathbf{0}\}$ . Then there are *no* vectors  $\mathbf{x}_i$  in the above proof, and the argument shows that if  $\{\mathbf{u}_1, \dots, \mathbf{u}_m\}$  and  $\{\mathbf{w}_1, \dots, \mathbf{w}_p\}$  are bases of  $U$  and  $W$  respectively, then  $\{\mathbf{u}_1, \dots, \mathbf{u}_m, \mathbf{w}_1, \dots, \mathbf{w}_p\}$  is a basis of  $U + W$ . In this case  $U + W$  is said to be a **direct sum** (written  $U \oplus W$ ); we return to this in Chapter ??.

## Exercises for 6.4

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**Exercise 6.4.1** In each case, find a basis for  $V$  that includes the vector  $\mathbf{v}$ .

a.  $V = \mathbb{R}^3$ ,  $\mathbf{v} = (1, -1, 1)$

b.  $V = \mathbb{R}^3$ ,  $\mathbf{v} = (0, 1, 1)$

c.  $V = \mathbf{M}_{22}$ ,  $\mathbf{v} = \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}$

d.  $V = \mathbf{P}_2$ ,  $\mathbf{v} = x^2 - x + 1$

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b.  $\{(0, 1, 1), (1, 0, 0), (0, 1, 0)\}$

d.  $\{x^2 - x + 1, 1, x\}$

**Exercise 6.4.2** In each case, find a basis for  $V$  among the given vectors.

a.  $V = \mathbb{R}^3$ ,  
 $\{(1, 1, -1), (2, 0, 1), (-1, 1, -2), (1, 2, 1)\}$

b.  $V = \mathbf{P}_2$ ,  $\{x^2 + 3, x + 2, x^2 - 2x - 1, x^2 + x\}$

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b. Any three except  $\{x^2 + 3, x + 2, x^2 - 2x - 1\}$

**Exercise 6.4.3** In each case, find a basis for  $V$  containing  $\mathbf{v}$  and  $\mathbf{w}$ .

a.  $V = \mathbb{R}^4$ ,  $\mathbf{v} = (1, -1, 1, -1)$ ,  $\mathbf{w} = (0, 1, 0, 1)$

b.  $V = \mathbb{R}^4$ ,  $\mathbf{v} = (0, 0, 1, 1)$ ,  $\mathbf{w} = (1, 1, 1, 1)$

c.  $V = \mathbf{M}_{22}$ ,  $\mathbf{v} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$ ,  $\mathbf{w} = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$

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

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b. Add  $(0, 1, 0, 0)$  and  $(0, 0, 1, 0)$ .

d. Add 1 and  $x^3$ .

**Exercise 6.4.4**

- a. If  $z$  is not a real number, show that  $\{z, z^2\}$  is a basis of the real vector space  $\mathbb{C}$  of all complex numbers.
- b. If  $z$  is neither real nor pure imaginary, show that  $\{z, \bar{z}\}$  is a basis of  $\mathbb{C}$ .

- b. If  $z = a + bi$ , then  $a \neq 0$  and  $b \neq 0$ . If  $rz + s\bar{z} = 0$ , then  $(r + s)a = 0$  and  $(r - s)b = 0$ . This means that  $r + s = 0 = r - s$ , so  $r = s = 0$ . Thus  $\{z, \bar{z}\}$  is independent; it is a basis because  $\dim \mathbb{C} = 2$ .

**Exercise 6.4.5** In each case use Theorem 6.4.4 to decide if  $S$  is a basis of  $V$ .

- a.  $V = \mathbf{M}_{22}$ ;  
 $S = \left\{ \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}, \begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 1 & 1 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} \right\}$
- b.  $V = \mathbf{P}_3$ ;  $S = \{2x^2, 1 + x, 3, 1 + x + x^2 + x^3\}$

- b. The polynomials in  $S$  have distinct degrees.

**Exercise 6.4.6**

- a. Find a basis of  $\mathbf{M}_{22}$  consisting of matrices with the property that  $A^2 = A$ .
- b. Find a basis of  $\mathbf{P}_3$  consisting of polynomials whose coefficients sum to 4. What if they sum to 0?

- b.  $\{4, 4x, 4x^2, 4x^3\}$  is one such basis of  $\mathbf{P}_3$ . However, there is *no* basis of  $\mathbf{P}_3$  consisting of polynomials that have the property that their coefficients sum to zero. For if such a basis exists, then every polynomial in  $\mathbf{P}_3$  would have this property (because sums and scalar multiples of such polynomials have the same property).

**Exercise 6.4.7** If  $\{\mathbf{u}, \mathbf{v}, \mathbf{w}\}$  is a basis of  $V$ , determine which of the following are bases.

- a.  $\{\mathbf{u} + \mathbf{v}, \mathbf{u} + \mathbf{w}, \mathbf{v} + \mathbf{w}\}$
- b.  $\{2\mathbf{u} + \mathbf{v} + 3\mathbf{w}, 3\mathbf{u} + \mathbf{v} - \mathbf{w}, \mathbf{u} - 4\mathbf{w}\}$
- c.  $\{\mathbf{u}, \mathbf{u} + \mathbf{v} + \mathbf{w}\}$
- d.  $\{\mathbf{u}, \mathbf{u} + \mathbf{w}, \mathbf{u} - \mathbf{w}, \mathbf{v} + \mathbf{w}\}$

- b. Not a basis.
- d. Not a basis.

**Exercise 6.4.8**

- a. Can two vectors span  $\mathbb{R}^3$ ? Can they be linearly independent? Explain.
- b. Can four vectors span  $\mathbb{R}^3$ ? Can they be linearly independent? Explain.

- b. Yes; no.

**Exercise 6.4.9** Show that any nonzero vector in a finite dimensional vector space is part of a basis.

**Exercise 6.4.10** If  $A$  is a square matrix, show that  $\det A = 0$  if and only if some row is a linear combination of the others. \_\_\_\_\_  
 $\det A = 0$  if and only if  $A$  is not invertible; if and only if the rows of  $A$  are dependent (Theorem 5.2.3); if and only if some row is a linear combination of the others (Lemma 6.4.2).

**Exercise 6.4.11** Let  $D$ ,  $I$ , and  $X$  denote finite, nonempty sets of vectors in a vector space  $V$ . Assume that  $D$  is dependent and  $I$  is independent. In each case answer yes or no, and defend your answer.

- a. If  $X \supseteq D$ , must  $X$  be dependent?
- b. If  $X \subseteq D$ , must  $X$  be dependent?
- c. If  $X \supseteq I$ , must  $X$  be independent?
- d. If  $X \subseteq I$ , must  $X$  be independent?

b. No.  $\{(0, 1), (1, 0)\} \subseteq \{(0, 1), (1, 0), (1, 1)\}$ .

d. Yes. See Exercise 6.3.15.

**Exercise 6.4.12** If  $U$  and  $W$  are subspaces of  $V$  and  $\dim U = 2$ , show that either  $U \subseteq W$  or  $\dim(U \cap W) \leq 1$ .

**Exercise 6.4.13** Let  $A$  be a nonzero  $2 \times 2$  matrix and write  $U = \{X \text{ in } \mathbf{M}_{22} \mid XA = AX\}$ . Show that  $\dim U \geq 2$ . [Hint:  $I$  and  $A$  are in  $U$ .]

**Exercise 6.4.14** If  $U \subseteq \mathbb{R}^2$  is a subspace, show that  $U = \{0\}$ ,  $U = \mathbb{R}^2$ , or  $U$  is a line through the origin.

**Exercise 6.4.15** Given  $\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3, \dots, \mathbf{v}_k$ , and  $\mathbf{v}$ , let  $U = \text{span}\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k\}$  and  $W = \text{span}\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k, \mathbf{v}\}$ . Show that either  $\dim W = \dim U$  or  $\dim W = 1 + \dim U$ .

If  $\mathbf{v} \in U$  then  $W = U$ ; if  $\mathbf{v} \notin U$  then  $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k, \mathbf{v}\}$  is a basis of  $W$  by the independent lemma.

**Exercise 6.4.16** Suppose  $U$  is a subspace of  $\mathbf{P}_1$ ,  $U \neq \{0\}$ , and  $U \neq \mathbf{P}_1$ . Show that either  $U = \mathbb{R}$  or  $U = \mathbb{R}(a+x)$  for some  $a$  in  $\mathbb{R}$ .

**Exercise 6.4.17** Let  $U$  be a subspace of  $V$  and assume  $\dim V = 4$  and  $\dim U = 2$ . Does every basis of  $V$  result from adding (two) vectors to some basis of  $U$ ? Defend your answer.

**Exercise 6.4.18** Let  $U$  and  $W$  be subspaces of a vector space  $V$ .

a. If  $\dim V = 3$ ,  $\dim U = \dim W = 2$ , and  $U \neq W$ , show that  $\dim(U \cap W) = 1$ .

b. Interpret (a.) geometrically if  $V = \mathbb{R}^3$ .

b. Two distinct planes through the origin ( $U$  and  $W$ ) meet in a line through the origin ( $U \cap W$ ).

**Exercise 6.4.19** Let  $U \subseteq W$  be subspaces of  $V$  with  $\dim U = k$  and  $\dim W = m$ , where  $k < m$ . If  $k < l < m$ , show that a subspace  $X$  exists where  $U \subseteq X \subseteq W$  and  $\dim X = l$ .

**Exercise 6.4.20** Let  $B = \{\mathbf{v}_1, \dots, \mathbf{v}_n\}$  be a *maximal* independent set in a vector space  $V$ . That is, no set of more than  $n$  vectors  $S$  is independent. Show that  $B$  is a basis of  $V$ .

**Exercise 6.4.21** Let  $B = \{\mathbf{v}_1, \dots, \mathbf{v}_n\}$  be a *minimal* spanning set for a vector space  $V$ . That is,  $V$  cannot be spanned by fewer than  $n$  vectors. Show that  $B$  is a basis of  $V$ .

**Exercise 6.4.22**

a. Let  $p(x)$  and  $q(x)$  lie in  $\mathbf{P}_1$  and suppose that  $p(1) \neq 0$ ,  $q(2) \neq 0$ , and  $p(2) = 0 = q(1)$ . Show that  $\{p(x), q(x)\}$  is a basis of  $\mathbf{P}_1$ . [Hint: If  $rp(x) + sq(x) = 0$ , evaluate at  $x = 1$ ,  $x = 2$ .]

b. Let  $B = \{p_0(x), p_1(x), \dots, p_n(x)\}$  be a set of polynomials in  $\mathbf{P}_n$ . Assume that there exist numbers  $a_0, a_1, \dots, a_n$  such that  $p_i(a_i) \neq 0$  for each  $i$  but  $p_i(a_j) = 0$  if  $i$  is different from  $j$ . Show that  $B$  is a basis of  $\mathbf{P}_n$ .

**Exercise 6.4.23** Let  $V$  be the set of all infinite sequences  $(a_0, a_1, a_2, \dots)$  of real numbers. Define addition and scalar multiplication by

$$(a_0, a_1, \dots) + (b_0, b_1, \dots) = (a_0 + b_0, a_1 + b_1, \dots)$$

and

$$r(a_0, a_1, \dots) = (ra_0, ra_1, \dots)$$

a. Show that  $V$  is a vector space.

b. Show that  $V$  is not finite dimensional.

c. [For those with some calculus.] Show that the set of convergent sequences (that is,  $\lim_{n \rightarrow \infty} a_n$  exists) is a subspace, also of infinite dimension.

b. The set  $\{(1, 0, 0, 0, \dots), (0, 1, 0, 0, 0, \dots), (0, 0, 1, 0, 0, \dots), \dots\}$  contains independent subsets of arbitrary size.

**Exercise 6.4.24** Let  $A$  be an  $n \times n$  matrix of rank  $r$ . If  $U = \{X \text{ in } \mathbf{M}_n \mid AX = 0\}$ , show that  $\dim U = n(n - r)$ . [Hint: Exercise 6.3.34.]

**Exercise 6.4.25** Let  $U$  and  $W$  be subspaces of  $V$ .

a. Show that  $U + W$  is a subspace of  $V$  containing both  $U$  and  $W$ .

b. Show that  $\text{span}\{\mathbf{u}, \mathbf{w}\} = \mathbb{R}\mathbf{u} + \mathbb{R}\mathbf{w}$  for any vectors  $\mathbf{u}$  and  $\mathbf{w}$ .

c. Show that

$$\begin{aligned} & \text{span}\{\mathbf{u}_1, \dots, \mathbf{u}_m, \mathbf{w}_1, \dots, \mathbf{w}_n\} \\ &= \text{span}\{\mathbf{u}_1, \dots, \mathbf{u}_m\} + \text{span}\{\mathbf{w}_1, \dots, \mathbf{w}_n\} \end{aligned}$$

for any vectors  $\mathbf{u}_i$  in  $U$  and  $\mathbf{w}_j$  in  $W$ .

$$\mathbb{R}\mathbf{u} + \mathbb{R}\mathbf{w} = \{r\mathbf{u} + s\mathbf{w} \mid r, s \text{ in } \mathbb{R}\} = \text{span}\{\mathbf{u}, \mathbf{w}\}$$

**Exercise 6.4.26** If  $A$  and  $B$  are  $m \times n$  matrices, show that  $\text{rank}(A + B) \leq \text{rank } A + \text{rank } B$ . [*Hint*: If  $U$  and  $V$  are the column spaces of  $A$  and  $B$ , respectively, show that the column space of  $A + B$  is contained in  $U + V$  and that  $\dim(U + V) \leq \dim U + \dim V$ . (See Theorem 6.4.5.)]

