

**3.2.28** The three column vectors are linearly independent, since  $\text{rref} \begin{bmatrix} 1 & 1 & 1 \\ 1 & 2 & 5 \\ 1 & 3 & 7 \end{bmatrix} = I_3$ .

Therefore, the three columns form a basis of  $\text{im}(A)(= \mathbb{R}^3)$ :

$$\begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}, \begin{bmatrix} 1 \\ 5 \\ 7 \end{bmatrix}.$$

Another sensible choice for a basis of  $\text{im}(A)$  is  $\vec{e}_1, \vec{e}_2, \vec{e}_3$ .

**3.2.29** The three column vectors of  $A$  span all of  $\mathbb{R}^2$ , so that  $\text{im}(A) = \mathbb{R}^2$ . We can choose any two of the columns of  $A$  to form a basis of  $\text{im}(A)$ ; another sensible choice is  $\vec{e}_1, \vec{e}_2$ .

**3.2.30**  $\text{im}(A) = \text{span}(\vec{e}_1, \vec{e}_2)$

We can choose  $\vec{e}_1, \vec{e}_2$  as a basis of  $\text{im}(A)$ .

3.2.38 a Using the terminology introduced in the exercise, we need to show that any vector  $\vec{v}$  in  $V$  is a linear combination of  $\vec{v}_1, \dots, \vec{v}_m$ . Choose a specific vector  $\vec{v}$  in  $V$ . Since we can find no more than  $m$  linearly independent vectors in  $V$ , the  $m + 1$  vectors  $\vec{v}_1, \dots, \vec{v}_m, \vec{v}$  will be linearly dependent. Since the vectors  $\vec{v}_1, \dots, \vec{v}_m$  are independent,  $\vec{v}$  must be redundant, meaning that  $\vec{v}$  is a linear combination of  $\vec{v}_1, \dots, \vec{v}_m$ , as claimed.

b With the terminology introduced in part a, we can let  $V = \text{im}[\vec{v}_1 \ \cdots \ \vec{v}_m]$ .

3.2.39 Yes; the vectors are linearly independent. The vectors in the list  $\vec{v}_1, \dots, \vec{v}_m$  are linearly independent (and therefore non-redundant), and  $\vec{v}$  is non-redundant since it fails to be in the span of  $\vec{v}_1, \dots, \vec{v}_m$ .

3.2.40 Yes; by Theorem 3.2.8,  $\ker(A) = \{\vec{0}\}$  and  $\ker(B) = \{\vec{0}\}$ . Then  $\ker(AB) = \{\vec{0}\}$  by Exercise 3.1.51, so that the columns of  $AB$  are linearly independent, by Theorem 3.2.8.

3.2.41 To show that the columns of  $B$  are linearly independent, we show that  $\ker(B) = \{\vec{0}\}$ . Indeed, if  $B\vec{x} = \vec{0}$ , then  $AB\vec{x} = A\vec{0} = \vec{0}$ , so that  $\vec{x} = \vec{0}$  (since  $AB = I_m$ ).

By Theorem 3.2.8,  $\text{rank}(B) = \# \text{ columns} = m$ , so that  $m \leq n$  and in fact  $m < n$  (we are told that  $m \neq n$ ). This implies that the rank of the  $m \times n$  matrix  $A$  is less than  $n$ , so that the columns of  $A$  are linearly dependent (by Theorem 3.2.8).

3.2.42 We can use the hint and form the dot product of  $\vec{v}_i$  and both sides of the relation

$$c_1\vec{v}_1 + \cdots + c_i\vec{v}_i + \cdots + c_m\vec{v}_m = \vec{0}:$$

$$(c_1\vec{v}_1 + \cdots + c_i\vec{v}_i + \cdots + c_m\vec{v}_m) \cdot \vec{v}_i = \vec{0} \cdot \vec{v}_i, \text{ so that } c_1(\vec{v}_1 \cdot \vec{v}_i) + \cdots + c_i(\vec{v}_i \cdot \vec{v}_i) + \cdots + c_m(\vec{v}_m \cdot \vec{v}_i) = 0.$$

Since  $\vec{v}_i$  is perpendicular to all the other  $\vec{v}_j$ , we will have  $\vec{v}_i \cdot \vec{v}_j = 0$  whenever  $j \neq i$ ; since  $\vec{v}_i$  is a unit vector, we will have  $\vec{v}_i \cdot \vec{v}_i = 1$ . Therefore, the equation above simplifies to  $c_i = 0$ .

Since this reasoning applies to all  $i = 1, \dots, m$ , we have only the trivial relation among the vectors  $\vec{v}_1, \vec{v}_2, \dots, \vec{v}_m$ , so that these vectors are linearly independent, as claimed.

3.2.43 Consider a linear relation  $c_1\vec{v}_1 + c_2(\vec{v}_1 + \vec{v}_2) + c_3(\vec{v}_1 + \vec{v}_2 + \vec{v}_3) = \vec{0}$ , or,  $(c_1 + c_2 + c_3)\vec{v}_1 + (c_2 + c_3)\vec{v}_2 + c_3\vec{v}_3 = \vec{0}$ . Since there is only the trivial relation among the vectors  $\vec{v}_1, \vec{v}_2, \vec{v}_3$ , we must have  $c_1 + c_2 + c_3 = c_2 + c_3 = c_3 = 0$ , so that  $c_3 = 0$  and then  $c_2 = 0$  and then  $c_1 = 0$ , as claimed.

3.2.44 Yes; this is a special case of Exercise 40 (recall that  $\ker(A) = \{\vec{0}\}$ , by Theorem 3.1.7b).

3.2.45 Yes; if  $A$  is invertible, then  $\ker(A) = \{\vec{0}\}$ , so that the columns of  $A$  are linearly independent, by Theorem 3.2.8.

3.2.46 Solve the system 
$$\begin{bmatrix} x_1 & + & 2x_2 & & + & 3x_4 & + & 5x_5 & = & 0 \\ & & & x_3 & + & 4x_4 & + & 6x_5 & = & 0 \end{bmatrix}.$$

The solutions are of the form

**3.3.22**  $\text{rref} \begin{bmatrix} 2 & 4 & 8 \\ 4 & 5 & 1 \\ 7 & 9 & 3 \end{bmatrix} = \begin{bmatrix} 1 & 0 & -6 \\ 0 & 1 & 5 \\ 0 & 0 & 0 \end{bmatrix}$ . It is clear that the third vector is redundant, and we quickly see that the

vector  $\begin{bmatrix} -6 \\ 5 \\ -1 \end{bmatrix}$  is in the kernel. Since this is the only redundant column,  $\left( \begin{bmatrix} -6 \\ 5 \\ -1 \end{bmatrix} \right)$  is a basis of the kernel. Thus,

a basis of  $\text{im}(A)$  is  $\left( \begin{bmatrix} 2 \\ 4 \\ 7 \end{bmatrix}, \begin{bmatrix} 4 \\ 5 \\ 9 \end{bmatrix} \right)$ .

3.3.26 a We notice that each of the six matrices has two identical columns. In matrices  $C$  and  $L$ , the second column is identical to the third, so that  $\ker(C) = \ker(L) = \text{span} \begin{bmatrix} 0 \\ 1 \\ -1 \end{bmatrix}$ . In matrices  $H, T, X$  and  $Y$ , the first column is identical to the third, so that  $\ker(H) = \ker(T) = \ker(X) = \ker(Y) = \text{span} \begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix}$ . Thus, only  $L$  has the same kernel as  $C$ .

b We observe that each of the six matrices in the list has two identical rows. For example, the first and the last row of matrix  $C$  are identical, so that any vector  $\begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix}$  in  $\text{im}(C)$  will satisfy the equation  $y_1 = y_3$ . We can conclude that  $\text{im}(C) = \text{im}(H) = \text{im}(X) = \left\{ \begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix} : y_1 = y_3 \right\}$ ,  $\text{im}(L) = \left\{ \begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix} : y_1 = y_2 \right\}$ , and  $\text{im}(T) = \text{im}(Y) = \left\{ \begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix} : y_2 = y_3 \right\}$ .

c Our discussion in part *b* shows that the answer is matrix  $L$ .

**3.3.28** Form a  $4 \times 4$  matrix  $A$  with the given vectors as its columns. The matrix  $A$  reduces to

$$\begin{bmatrix} 1 & 0 & 0 & 2 \\ 0 & 1 & 0 & 3 \\ 0 & 0 & 1 & 4 \\ 0 & 0 & 0 & k-29 \end{bmatrix}.$$

This matrix can be reduced further to  $I_4$  if (and only if)  $k - 29 \neq 0$ , that is, if  $k \neq 29$ .

By Summary 3.3.10, the four given vectors form a basis of  $\mathbb{R}^4$  unless  $k = 29$ .

**3.3.29**  $x_1 = -\frac{3}{2}x_2 - \frac{1}{2}x_3$ ; let  $x_2 = s$  and  $x_3 = t$ . Then the solutions are of the form

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} -\frac{3}{2}s - \frac{1}{2}t \\ s \\ t \end{bmatrix} = s \begin{bmatrix} -\frac{3}{2} \\ 1 \\ 0 \end{bmatrix} + t \begin{bmatrix} -\frac{1}{2} \\ 0 \\ 1 \end{bmatrix}.$$

Multiplying the two vectors by 2 to simplify, we obtain the basis  $\begin{bmatrix} -3 \\ 2 \\ 0 \end{bmatrix}, \begin{bmatrix} -1 \\ 0 \\ 2 \end{bmatrix}$ .

**3.3.30** Proceeding as in Exercise 29, we find the basis  $\begin{bmatrix} 1 \\ 2 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} -1 \\ 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} -2 \\ 0 \\ 0 \\ 1 \end{bmatrix}$ .

**3.3.31** Proceeding as in Exercise 29, we can find the following basis of  $V$ :  $\begin{bmatrix} 1 \\ 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} -2 \\ 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} -4 \\ 0 \\ 0 \\ 1 \end{bmatrix}$ .

Now let  $A$  be the  $4 \times 3$  matrix with these three vectors as its columns. Then  $\text{im}(A) = V$  by Theorem 3.1.3, and  $\ker(A) = \{\vec{0}\}$  by Theorem 3.2.8, so that  $A$  does the job.

$$A = \begin{bmatrix} 1 & -2 & -4 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}.$$

**3.3.32** We need to find all vectors  $\vec{x}$  in  $\mathbb{R}^4$  such that  $\begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ -1 \\ 1 \end{bmatrix} = 0$  and  $\begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 2 \\ 3 \end{bmatrix} = 0$ .

This amounts to solving the system  $\begin{bmatrix} x_1 & & -x_3 & +x_4 & = & 0 \\ & x_2 & +2x_3 & +3x_4 & = & 0 \end{bmatrix}$ , which in turn amounts to finding the kernel of  $\begin{bmatrix} 1 & 0 & -1 & 1 \\ 0 & 1 & 2 & 3 \end{bmatrix}$ .

Using Kyle Numbers, we find the basis  $\begin{bmatrix} 1 \\ -2 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} -1 \\ -3 \\ 0 \\ 1 \end{bmatrix}$ .

**3.3.33** We can write  $V = \ker(A)$ , where  $A$  is the  $1 \times n$  matrix  $A = [c_1 \ c_2 \ \cdots \ c_n]$ .

Since at least one of the  $c_i$  is nonzero, the rank of  $A$  is 1, so that  $\dim(V) = \dim(\ker(A)) = n - \text{rank}(A) = n - 1$ , by Theorem 3.3.7.

A “hyperplane” in  $\mathbb{R}^2$  is a line, and a “hyperplane” in  $\mathbb{R}^3$  is just a plane.

**3.3.34** We can write  $V = \ker(A)$ , where  $A$  is the  $n \times m$  matrix with entries  $a_{ij}$ . Note that  $\text{rank}(A) \leq n$ . Therefore,  $\dim(V) = \dim(\ker(A)) = m - \text{rank}(A) \geq m - n$ , by Theorem 3.3.7.

**3.3.35** We need to find all vectors  $\vec{x}$  in  $\mathbb{R}^n$  such that  $\vec{v} \cdot \vec{x} = 0$ , or  $v_1x_1 + v_2x_2 + \cdots + v_nx_n = 0$ , where the  $v_i$  are the components of the vector  $\vec{v}$ . These vectors form a hyperplane in  $\mathbb{R}^n$  (see Exercise 33), so that the dimension of the space is  $n - 1$ .

**3.3.36** No; if  $\text{im}(A) = \ker(A)$  for an  $n \times n$  matrix  $A$ , then  $n = \dim(\ker(A)) + \dim(\text{im}(A)) = 2 \dim(\text{im}(A))$ , so that  $n$  is an even number.

**3.3.37** Since  $\dim(\ker(A)) = 5 - \text{rank}(A)$ , any  $4 \times 5$  matrix with rank 2 will do; for example,

$$A = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}.$$

**3.3.38 a** The rank of a  $3 \times 5$  matrix  $A$  is 0,1,2, or 3, so that  $\dim(\ker(A)) = 5 - \text{rank}(A)$  is 2,3,4, or 5.

b The rank of a  $7 \times 4$  matrix  $A$  is at most 4, so that  $\dim(\text{im}(A)) = \text{rank}(A)$  is 0,1,2,3, or 4.

**3.3.60** We can choose a basis  $\vec{v}_1, \dots, \vec{v}_p$  in  $V$ , where  $p = \dim(V)$ . Then  $\vec{v}_1, \dots, \vec{v}_p$  are linearly independent vectors in  $W$ , so that  $\dim(V) = p \leq \dim(W)$ , by Theorem 3.3.4a, as claimed.

**3.3.61** We can choose a basis  $\vec{v}_1, \dots, \vec{v}_p$  of  $V$ , where  $p = \dim(V) = \dim(W)$ . Then  $\vec{v}_1, \dots, \vec{v}_p$  is a basis of  $W$  as well, by Theorem 3.3.4c, so that  $V = W = \text{span}(\vec{v}_1, \dots, \vec{v}_p)$ , as claimed.

**3.3.62** Consider a basis  $\vec{v}_1, \dots, \vec{v}_n$  of  $V$ . Since the  $\vec{v}_i$  are  $n$  linearly independent vectors in  $\mathbb{R}^n$ , they form a basis of  $\mathbb{R}^n$  (by parts (vii) and (ix) of Summary 3.3.10), so that  $V = \text{span}(\vec{v}_1, \dots, \vec{v}_n) = \mathbb{R}^n$ , as claimed. (Note that Exercise 3.3.62 is a special case of Exercise 3.3.61.)

**3.3.63**  $\dim(V + W) = \dim(V) + \dim(W)$ , by Exercise 3.2.51b.

**3.3.64** Suppose that  $V \cap W = \{\vec{0}\}$  and  $\dim(V) + \dim(W) = n$ .

Choose a basis  $\vec{v}_1, \dots, \vec{v}_p$  of  $V$  and a basis  $\vec{w}_1, \dots, \vec{w}_q$  in  $W$ ; note that  $p + q = n$ . By Exercise 3.2.51b, the  $n$  vectors  $\vec{v}_1, \dots, \vec{v}_p, \vec{w}_1, \dots, \vec{w}_q$  in  $\mathbb{R}^n$  are linearly independent, so that they form a basis of  $\mathbb{R}^n$  (by parts (vii) and (ix) of Summary 3.3.10). By Theorem 3.2.10, any vector  $\vec{x}$  can be written uniquely as

$\vec{x} = (c_1\vec{v}_1 + \dots + c_p\vec{v}_p) + (d_1\vec{w}_1 + \dots + d_q\vec{w}_q)$ , with  $\vec{v} = c_1\vec{v}_1 + \dots + c_p\vec{v}_p$  in  $V$  and  $\vec{w} = d_1\vec{w}_1 + \dots + d_q\vec{w}_q$  in  $W$ , which gives the desired representation.

Conversely, suppose  $V$  and  $W$  are complements. Let us first show that  $V \cap W = \{\vec{0}\}$  in this case.

Indeed, if  $\vec{x}$  is in  $V \cap W$ , then we can write  $\vec{x} = \vec{x} + \vec{0} = \vec{0} + \vec{x}$

$$\begin{array}{cccc} \uparrow & \uparrow & \uparrow & \uparrow \\ \text{in} & \text{in} & \text{in} & \text{in} \\ V & W & V & W \end{array}$$

Since this representation is unique (by definition of complements), we must have  $\vec{x} = \vec{0}$ , so that  $V \cap W = \{\vec{0}\}$ . By definition of complements, we have  $\mathbb{R}^n = V + W$ , so that  $n = \dim(V + W) = \dim(V) + \dim(W)$ , by Exercise 63.

**3.3.65** Note that  $\text{im}(A) = \text{span}(\vec{v}_1, \dots, \vec{v}_p, \vec{w}_1, \dots, \vec{w}_q) = V$ , since the  $\vec{w}_j$  alone span  $V$ .

To find a basis of  $V = \text{im}(A)$ , we omit the redundant vectors from the list  $\vec{v}_1, \dots, \vec{v}_p, \vec{w}_1, \dots, \vec{w}_q$ , by Theorem 3.2.4. Since the vectors  $\vec{v}_1, \dots, \vec{v}_p$  are linearly independent, none of them are redundant, so that our basis of  $V$  contains all vectors  $\vec{v}_1, \dots, \vec{v}_p$  and some of the vectors from the list  $\vec{w}_1, \dots, \vec{w}_q$ .

**3.3.66** Use Exercise 65 with  $\vec{v}_1 = \begin{bmatrix} 1 \\ 2 \\ 3 \\ 4 \end{bmatrix}$ ,  $\vec{v}_2 = \begin{bmatrix} 1 \\ 4 \\ 6 \\ 8 \end{bmatrix}$ , and  $\vec{w}_i = \vec{e}_i$  for  $i = 1, 2, 3, 4$ .

$$\text{Now rref} \begin{bmatrix} 1 & 1 & 1 & 0 & 0 & 0 \\ 2 & 4 & 0 & 1 & 0 & 0 \\ 3 & 6 & 0 & 0 & 1 & 0 \\ 4 & 8 & 0 & 0 & 0 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 2 & 0 & 0 & -\frac{1}{4} \\ 0 & 1 & -1 & 0 & 0 & \frac{1}{4} \\ 0 & 0 & 0 & 1 & 0 & -\frac{1}{2} \\ 0 & 0 & 0 & 0 & 1 & -\frac{3}{4} \end{bmatrix}.$$

Picking the non-redundant columns gives the basis  $\begin{bmatrix} 1 \\ 2 \\ 3 \\ 4 \end{bmatrix}, \begin{bmatrix} 1 \\ 4 \\ 6 \\ 8 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \\ 0 \end{bmatrix}$ .

**3.3.82** Write  $A = [\vec{v}_1 \ \dots \ \vec{v}_m]$  and  $B = [\vec{w}_1 \ \dots \ \vec{w}_m]$ , so that  $A + B = [\vec{v}_1 + \vec{w}_1 \ \dots \ \vec{v}_m + \vec{w}_m]$ . Any linear combination of the columns of  $A + B$ ,  $\vec{y} = c_1(\vec{v}_1 + \vec{w}_1) + \dots + c_m(\vec{v}_m + \vec{w}_m)$ , can be written as

$$\vec{y} = \underbrace{(c_1\vec{v}_1 + \dots + c_m\vec{v}_m)}_{\text{in im}(A)} + \underbrace{(c_1\vec{w}_1 + \dots + c_m\vec{w}_m)}_{\text{in im}(B)}$$

so that  $\text{im}(A + B) \subseteq \text{im}(A) + \text{im}(B)$  (see Exercise 3.2.50). Since  $\dim(V + W) \leq \dim(V) + \dim(W)$ , by Exercise 3.3.67, we can conclude that  $\text{rank}(A+B) = \dim(\text{im}(A+B)) \leq \dim(\text{im}(A)) + \dim(\text{im}(B)) = \text{rank}(A) + \text{rank}(B)$ .

Summary:  $\text{rank}(A + B) \leq \text{rank}(A) + \text{rank}(B)$ .