

# MA1522 Linear Algebra for Computing

## Lecture 6: Span and Linear Dependence/Independence

Yang Yue

Department of Mathematics  
National University of Singapore

17 February, 2025

# Outline

Questions posed in Dr.Teo's Lectures

Challenges about Optional Topic: Abstract Vector Spaces

## Question one in Section 3.3

Let  $\mathbf{u}_1 = \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix}$ ,  $\mathbf{u}_2 = \begin{pmatrix} 0 \\ 1 \\ -1 \end{pmatrix}$ ,  $\mathbf{u}_3 = \begin{pmatrix} 2 \\ 1 \\ 1 \end{pmatrix}$ , and  $\mathbf{v} = \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}$ .

- (i) Is  $\mathbf{v}$  in  $\text{span}\{\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3\}$ ?
- (ii) If it is, write  $\mathbf{v}$  as a linear combination of  $\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3$ ,

$$\mathbf{v} = c_1\mathbf{u}_1 + c_2\mathbf{u}_2 + c_3\mathbf{u}_3.$$

- (iii) Are the coefficients  $c_1, c_2, c_3$  unique?

Find a vector  $\begin{pmatrix} x \\ y \\ z \end{pmatrix}$  that is not in  $\text{span}\{\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3\}$ .

## Slide 24: Linear Combinations

### Definition

Let  $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k$  be vectors in  $\mathbb{R}^n$ . A linear combination of the vectors  $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k$  is

$$c_1\mathbf{u}_1 + c_2\mathbf{u}_2 + \cdots + c_k\mathbf{u}_k,$$

for some  $c_1, c_2, \dots, c_k \in \mathbb{R}$ . The scalars  $c_1, c_2, \dots, c_k$  are called coefficients.

## Slide 26: Linear Span

### Definition

Let  $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k$  be vectors in  $\mathbb{R}^n$ . The span of  $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k$  is the subset of  $\mathbb{R}^n$  containing **all the linear combinations** of  $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k$ ,

$$\text{span}\{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k\} = \{ c_1\mathbf{u}_1 + c_2\mathbf{u}_2 + \cdots + c_k\mathbf{u}_k \mid c_1, c_2, \dots, c_k \in \mathbb{R} \}.$$

We also define the span of the empty set  $\text{span } \emptyset = \{\mathbf{0}\}$ .

That is every vector  $\mathbf{v}$  in the set  $\text{span}\{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k\}$  is a linear combination of  $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k$ ,

$$\mathbf{v} = c_1\mathbf{u}_1 + c_2\mathbf{u}_2 + \cdots + c_k\mathbf{u}_k,$$

for some scalars  $c_1, c_2, \dots, c_k$ .

## Slide 31: Algorithm to Check for Linear Combination

Let  $S = \{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k\}$  be a set of vectors in  $\mathbb{R}^n$ .

- ▶ Form the  $n \times k$  matrix  $\mathbf{A} = \begin{pmatrix} \mathbf{u}_1 & \mathbf{u}_2 & \cdots & \mathbf{u}_k \end{pmatrix}$  whose columns are the vectors in  $S$ .
- ▶ Then a vector  $\mathbf{v}$  in  $\mathbb{R}^n$  is in  $\text{span}\{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k\}$  if and only if the system  $\mathbf{A}\mathbf{x} = \mathbf{v}$  is consistent.
- ▶ If the system is consistent, then the solutions to the system are the possible coefficients of the linear combination. That

is, if  $\mathbf{u} = \begin{pmatrix} c_1 \\ c_2 \\ \vdots \\ c_k \end{pmatrix}$  is a solution to  $\mathbf{A}\mathbf{x} = \mathbf{v}$ , then

$$\mathbf{v} = c_1\mathbf{u}_1 + c_2\mathbf{u}_2 + \cdots + c_k\mathbf{u}_k.$$

Explicitly,  $\mathbf{v} \in \text{span}\{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k\}$  if and only if  
 $(\begin{array}{cccc|c} \mathbf{u}_1 & \mathbf{u}_2 & \cdots & \mathbf{u}_k & \mathbf{v} \end{array})$  is consistent.

## Answer to Question one in Section 3.3, part 1

Q: Let  $\mathbf{u}_1 = \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix}$ ,  $\mathbf{u}_2 = \begin{pmatrix} 0 \\ 1 \\ -1 \end{pmatrix}$ ,  $\mathbf{u}_3 = \begin{pmatrix} 2 \\ 1 \\ 1 \end{pmatrix}$ , and  $\mathbf{v} = \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}$ . Is  $\mathbf{v}$  in  $\text{span}\{\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3\}$ ?

Answer: By the algorithm on Slide 31, we form the augmented matrix:

$$\left( \begin{array}{ccc|c} 1 & 0 & 2 & 1 \\ 0 & 1 & 1 & 1 \\ 1 & -1 & 1 & 0 \end{array} \right) \xrightarrow{R_3 - R_1} \xrightarrow{R_3 + R_2} \left( \begin{array}{ccc|c} 1 & 0 & 2 & 1 \\ 0 & 1 & 1 & 1 \\ 0 & 0 & 0 & 0 \end{array} \right).$$

The system is consistent. Hence,  $\mathbf{v}$  is in  $\text{span}\{\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3\}$ .

## Answer to Question one in Section 3.3, part 1 (conti.)

Q: (following (i))

(ii) Write  $\mathbf{v}$  as a linear combination of  $\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3$ ,

$$\mathbf{v} = c_1\mathbf{u}_1 + c_2\mathbf{u}_2 + c_3\mathbf{u}_3.$$

(iii) Are the coefficients  $c_1, c_2, c_3$  unique?

Answer: We have had the REF form of the augmented matrix:

$$\left( \begin{array}{ccc|c} 1 & 0 & 2 & 1 \\ 0 & 1 & 1 & 1 \\ 0 & 0 & 0 & 0 \end{array} \right).$$

The system has solutions  $c_1 = 1 - 2s$ ,  $c_2 = 1 - s$  and  $c_3 = s$ , where  $s \in \mathbb{R}$ . We can let  $s = 0$ , and  $c_1 = c_2 = 1$  and  $c_3 = 0$ . Namely

$$\mathbf{v} = \mathbf{u}_1 + \mathbf{u}_2.$$

The answer is not unique, for example, we let  $s = 1$  and  $c_1 = -1$ ,  $c_2 = 0$  and  $c_3 = 1$ ,

$$\mathbf{v} = -\mathbf{u}_1 + \mathbf{u}_3.$$



## Answer to Question one in Section 3.2, part 2

Q: Let  $\mathbf{u}_1 = \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix}$ ,  $\mathbf{u}_2 = \begin{pmatrix} 0 \\ 1 \\ -1 \end{pmatrix}$ , and  $\mathbf{u}_3 = \begin{pmatrix} 2 \\ 1 \\ 1 \end{pmatrix}$ . Find a vector  $\begin{pmatrix} x \\ y \\ z \end{pmatrix}$  that is not in  $\text{span}\{\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3\}$ .

Answer: We can take  $\mathbf{v} = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$ , and form the augmented matrices

$$\left( \begin{array}{ccc|c} 1 & 0 & 2 & 1 \\ 0 & 1 & 1 & 1 \\ 1 & -1 & 1 & 1 \end{array} \right) \xrightarrow{R_3 - R_1} \xrightarrow{R_3 + R_2} \left( \begin{array}{ccc|c} 1 & 0 & 2 & 1 \\ 0 & 1 & 1 & 1 \\ 0 & 0 & 0 & 1 \end{array} \right).$$

The system is inconsistent. Thus  $\mathbf{v}$  is not in  $\text{span}\{\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3\}$ .

## Question Two in Section 3.3

Let  $S = \{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k\}$  be a set of  $k$  vectors in  $\mathbb{R}^n$ .

1. Show that if  $k < n$  then  $\text{span}(S) \neq \mathbb{R}^n$ .
2. If  $k > n$ , can we make any conclusion?

## Slide 37: Algorithm to check if $\text{span}(S) = \mathbb{R}^n$ .

Let  $S = \{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k\}$  be a set of vectors in  $\mathbb{R}^n$ .

- ▶ Form the  $n \times k$  matrix  $\mathbf{A} = \begin{pmatrix} \mathbf{u}_1 & \mathbf{u}_2 & \cdots & \mathbf{u}_k \end{pmatrix}$  whose columns are the vectors in  $S$ .
- ▶ Then  $\text{span}(S) = \mathbb{R}^n$  if and only if the system  $\mathbf{A}\mathbf{x} = \mathbf{v}$  is consistent for all  $\mathbf{v}$ .
- ▶ This is equivalent to the reduced row-echelon form of  $\mathbf{A}$  having no zero rows.

Explicitly,  $\text{span}\{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k\} = \mathbb{R}^n$  if and only if the reduced row-echelon form of  $\begin{pmatrix} \mathbf{u}_1 & \mathbf{u}_2 & \cdots & \mathbf{u}_k \end{pmatrix}$  has no zero rows.

## Answer to Question two in Section 3.2, part 1

Q: Let  $S = \{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k\}$  be a set of  $k$  vectors in  $\mathbb{R}^n$ . Show that if  $k < n$  then  $\text{span}(S) \neq \mathbb{R}^n$ .

Answer: By slide 37, we study the reduced row-echelon form of  $(\mathbf{u}_1 \ \mathbf{u}_2 \ \cdots \ \mathbf{u}_k)$ . Since  $k < n$ , the number of pivotal columns can be at most  $k$ , in other words, there are at most  $k$  many nonzero rows. Thus, there must be zero rows, because  $n > k$ . Therefore, we conclude  $\text{span}(S) \neq \mathbb{R}^n$ .

## Answer to Question two in Section 3.2, part 2

Let  $S = \{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k\}$  be a set of  $k$  vectors in  $\mathbb{R}^n$ . If  $k > n$ , can we make any conclusion?

Answer: We cannot make any conclusion. For example, let  $n = 2$ , and  $k = 3$ .

If we take  $\mathbf{u}_1 = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$ ,  $\mathbf{u}_2 = \begin{pmatrix} 0 \\ 1 \end{pmatrix}$  and  $\mathbf{u}_3 = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$ . Then  $\text{span}(S) = \mathbb{R}^2$  (because  $\mathbf{u}_1$  and  $\mathbf{u}_2$  already span  $\mathbb{R}^2$ ).

On the other hand, if we take  $\mathbf{u}_1 = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$ ,  $\mathbf{u}_2 = \begin{pmatrix} 2 \\ 0 \end{pmatrix}$  and

$\mathbf{u}_3 = \begin{pmatrix} 3 \\ 0 \end{pmatrix}$ . Then  $\text{span}(S) \neq \mathbb{R}^2$  (because the RREF has a zero row).

## Question one in Section 3.4

1. Show that the set containing the zero vector  $\{\mathbf{0}\}$  is a subspace.
2. Construct a set  $V$  such that it satisfies condition (i) and (ii) but not (iii); that is,  $V$  contains the origin and is closed under scalar multiplication, but not closed under addition.
3. Let  $V$  be a subspace of  $\mathbb{R}^n$  and  $S = \{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k\}$  a subset of  $V$ ,  $S \subseteq V$ . Show that the span of  $S$  is contained in  $V$ ,  $\text{span}(S) \subseteq V$ .

## Slide 58: Subspace

### Definition

A subset  $V$  of  $\mathbb{R}^n$  is a subspace if it satisfies the following properties.

- (i)  $V$  contains the zero vector  $\mathbf{0} \in V$ .
- (ii)  $V$  is closed under scalar multiplication. For any vector  $\mathbf{v}$  in  $V$  and scalar  $\alpha$ , the vector  $\alpha\mathbf{v}$  is in  $V$ .
- (iii)  $V$  is closed under addition. For any vectors  $\mathbf{u}, \mathbf{v}$  in  $V$ , the sum  $\mathbf{u} + \mathbf{v}$  is in  $V$ .

### Remark

- (1) Property (i) can be replaced with property (i'):  $V$  is **nonempty**.
- (2) Properties (ii) and (iii) is equivalent to property (ii'):  
For any  $\mathbf{u}, \mathbf{v}$  in  $V$ , and scalars  $\alpha, \beta$ , the linear combination  $\alpha\mathbf{u} + \beta\mathbf{v}$  is in  $V$ .

## Remarks

- ▶ Algebraic terminology: Let  $U$  be a set,  $f: U \rightarrow U$  a function/operation on  $U$ . A subset  $X$  of  $U$  is *closed under  $f$* , if for any  $x \in X$ ,  $f(x) \in X$ .
- ▶ Subspace is an instance of “substructure” in algebra.
- ▶ It turns out that for a subset  $V$  of the Euclidean space  $\mathbb{R}^n$  to satisfy all 10 axioms of being a vector space, suffice for it to satisfies only 3 of them.
- ▶ (This is because all other axioms are in “universal” form. If a “universal” property holds for a big set, then it holds for all its subsets.)



## Answer to Question one in Section 3.4, part 1

Q: Show that the set containing the zero vector  $Z = \{\mathbf{0}\}$  is a subspace.

Answer: We check the three properties:

- (i)  $V$  contains the zero vector  $\mathbf{0} \in V$ . Clearly,  $\mathbf{0} \in Z$ .
- (ii)  $V$  is closed under scalar multiplication. For any vector  $\mathbf{v}$  in  $V$  and scalar  $\alpha$ , the vector  $\alpha\mathbf{v}$  is in  $V$ .  
Only  $\mathbf{0}$  is in  $Z$ , and for any scalar  $\alpha$ ,  $\alpha\mathbf{0} = \mathbf{0} \in Z$ .
- (iii)  $V$  is closed under addition. For any vectors  $\mathbf{u}, \mathbf{v}$  in  $V$ , the sum  $\mathbf{u} + \mathbf{v}$  is in  $V$ .  
Again, we only have one sum  $\mathbf{0} + \mathbf{0} = \mathbf{0} \in Z$ .

## Answer to Question one in Section 3.4, part 2

Q: Construct a set  $V$  such that it satisfies condition (i) and (ii) but not (iii); that is,  $V$  contains the origin and is closed under scalar multiplication, but not closed under addition.

Answer: Consider the space  $\mathbb{R}^2$  and  $V$  to be the two axis. That is,

$$V = \{(x, 0) : x \in \mathbb{R}\} \cup \{(0, y) : y \in \mathbb{R}\}.$$

Then (i) and (ii) are satisfied, but (iii) failed because  $(1, 0) + (0, 1) = (1, 1) \notin V$ .

## Answer to Question one in Section 3.4, part 3

Q: Let  $V$  be a subspace of  $\mathbb{R}^n$  and  $S = \{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k\}$  a subset of  $V$ ,  $S \subseteq V$ . Show that the span of  $S$  is contained in  $V$ ,  $\text{span}(S) \subseteq V$ .

Answer: Recall that the elements  $\mathbf{w}$  in  $\text{span}(S)$  are linear combinations of  $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k$ , i.e., of the form

$$c_1\mathbf{u}_1 + \cdots + c_k\mathbf{u}_k$$

for some  $c_1, \dots, c_k \in \mathbb{R}$ .

Given  $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k \in V$ , since  $V$  is a subspace,  $V$  is closed under scalar multiplication, so  $c_1\mathbf{u}_1, \dots, c_k\mathbf{u}_k$  are in  $V$ .  $V$  is also closed under addition, because it is a subspace, we have

$$c_1\mathbf{u}_1 + \cdots + c_k\mathbf{u}_k \in V.$$

## Question two in Section 3.4

Is  $\mathbb{R}^2 \subseteq \mathbb{R}^3$ ?

Answer: No, because every element in  $\mathbb{R}^2$  has two coordinates, whereas  $\mathbb{R}^3$  has three.

That said,  $\mathbb{R}^2$  can be “embedded” into  $\mathbb{R}^3$  by  $(x, y) \mapsto (x, y, 0)$ . In other words, if we identify  $(x, y)$  with  $(x, y, 0)$ , then  $\mathbb{R}^2$  can be viewed as a subspace of  $\mathbb{R}^3$ .

## Question one in Section 3.5

Suppose  $\{\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3\}$  is linearly independent. Let

$$\mathbf{v}_1 = \mathbf{u}_1,$$

$$\mathbf{v}_2 = \mathbf{u}_1 + \mathbf{u}_2,$$

$$\mathbf{v}_3 = \mathbf{u}_1 + \mathbf{u}_2 + \mathbf{u}_3.$$

Show that  $\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$  is linearly independent too.

## Slide 82: Linearly Independent

### Definition

A set  $\{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k\}$  is linearly independent if the **only coefficients**  $c_1, c_2, \dots, c_k$  satisfying the equation

$$c_1\mathbf{u}_1 + c_2\mathbf{u}_2 + \cdots + c_k\mathbf{u}_k = \mathbf{0},$$

are  $c_1 = c_2 = \cdots = c_k = 0$ . Otherwise, we say that the set is linearly dependent.

# Remarks on Linear Independence

- ▶ In symbols,  $\{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k\}$  is linearly independent if for all  $c_1, \dots, c_k \in \mathbb{R}$ ,

$$c_1\mathbf{u}_1 + c_2\mathbf{u}_2 + \dots + c_k\mathbf{u}_k = \mathbf{0} \Rightarrow c_1 = c_2 = \dots = c_k = 0.$$

- ▶  $\{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k\}$  is linearly dependent if for some  $c_1, \dots, c_k \in \mathbb{R}$ , not all equal to 0 such that

$$c_1\mathbf{u}_1 + c_2\mathbf{u}_2 + \dots + c_k\mathbf{u}_k = \mathbf{0}.$$

- ▶ (It is a good exercise in logic to show the second statement is indeed a negation of the first.)

## Answer to Question one in Section 3.5

Suppose  $\{\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3\}$  is linearly independent. Let

$$\mathbf{v}_1 = \mathbf{u}_1,$$

$$\mathbf{v}_2 = \mathbf{u}_1 + \mathbf{u}_2,$$

$$\mathbf{v}_3 = \mathbf{u}_1 + \mathbf{u}_2 + \mathbf{u}_3.$$

Show that  $\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$  is linearly independent too.

Answer: Suppose that for some  $c_1, c_2, c_3 \in \mathbb{R}$  with

$$c_1\mathbf{v}_1 + c_2\mathbf{v}_2 + c_3\mathbf{v}_3 = \mathbf{0}.$$

We must show that  $c_1 = c_2 = c_3 = 0$ .



## Answer to Question one in Section 3.5 (conti.)

Substitute the  $\mathbf{u}$  vectors, we have

$$c_1\mathbf{u}_1 + c_2(\mathbf{u}_1 + \mathbf{u}_2) + c_3(\mathbf{u}_1 + \mathbf{u}_2 + \mathbf{u}_3) = \mathbf{0}.$$

Namely,

$$(c_1 + c_2 + c_3)\mathbf{u}_1 + (c_2 + c_3)\mathbf{u}_2 + c_3\mathbf{u}_3 = \mathbf{0}.$$

Since  $\{\mathbf{u}_1, \mathbf{u}_2, \mathbf{u}_3\}$  is linearly independent, we have

$$c_1 + c_2 + c_3 = c_2 + c_3 = c_3 = 0.$$

Hence  $c_1 = c_2 = c_3 = 0$ .

## Question two in Section 3.5

Let  $S = \{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k\}$  be a set of vectors in  $\mathbb{R}^n$ . Show that if  $k > n$ , then  $S$  is linearly dependent.

Answer: We need to show that for some scalars  $c_1, \dots, c_k \in \mathbb{R}$ , not all zero,  $c_1\mathbf{u}_1 + \dots + c_k\mathbf{u}_k = \mathbf{0}$ . In other words, the linear system with  $c_i$  as unknowns has nonzero solutions.

Form the augmented matrix, which is  $n \times k$ . Since  $k > n$ , it has infinitely many solutions (with at least  $k - n$  parameters). Thus  $S$  is linearly dependent.

## Challenge in Section 3.4

Prove that if a subset  $V$  of  $\mathbb{R}^n$  satisfies the 3 criteria of a subspace, then it satisfies all 10 axioms of a vector space.

Answer: By (ii) and (iii),  $V$  also equipped with the same addition and scalar multiplication.

Axioms 1,2,5,6,7,8 are in universal form, thus they also hold in  $V$ .

Axiom 3 holds by (i).

Axiom 4 holds because  $-\mathbf{u} = (-1)\mathbf{u} \in V$ .

# Definition of Abstract Vector Spaces

A set  $V$  equipped with **addition** and **scalar multiplication** is said to be a vector space over  $\mathbb{R}$  if it satisfies the following axioms.

1. (Commutative) For any vectors  $\mathbf{u}, \mathbf{v}$  in  $V$ ,  $\mathbf{u} + \mathbf{v} = \mathbf{v} + \mathbf{u}$ .
2. (Associative) For any vectors  $\mathbf{u}, \mathbf{v}, \mathbf{w}$  in  $V$ ,  
 $\mathbf{u} + (\mathbf{v} + \mathbf{w}) = (\mathbf{u} + \mathbf{v}) + \mathbf{w}$ .
3. (Zero vector) There is a vector  $\mathbf{0}$  in  $V$  such that  $\mathbf{0} + \mathbf{v} = \mathbf{v}$  for all vectors  $\mathbf{v}$  in  $V$ .
4. (Negative) For any vector  $\mathbf{u}$  in  $V$ , there exists a vector  $-\mathbf{u}$  in  $V$  such that  $\mathbf{u} + (-\mathbf{u}) = \mathbf{0}$ .
5. (Distribution) For any scalar  $a$  in  $\mathbb{R}$  and vectors  $\mathbf{u}, \mathbf{v}$  in  $V$ ,  
 $a(\mathbf{u} + \mathbf{v}) = a\mathbf{u} + a\mathbf{v}$ .
6. (Distribution) For any scalars  $a, b$  in  $\mathbb{R}$  and vector  $\mathbf{u}$  in  $V$ ,  
 $(a + b)\mathbf{u} = a\mathbf{u} + b\mathbf{u}$ .
7. (Associativity of scalar multiplication) For any scalars  $a, b$  in  $\mathbb{R}$  and vector  $\mathbf{u}$  in  $V$ ,  $a(b\mathbf{u}) = (ab)\mathbf{u}$ .
8. For any vector  $\mathbf{u}$  in  $V$ ,  $1\mathbf{u} = \mathbf{u}$ .

## Challenge in Section 3.3

Let  $S = \{\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_k\}$  be a set of vectors in  $\mathbb{R}^n$ . Referring to the properties of a spanning set or otherwise, show that the set  $V = \text{span}(S)$  is a (abstract) vector space. That is, it satisfies the 10 axioms of the definition of vector spaces.

Answer: By Challenge in Section 3.4, it suffices to show that  $V$  satisfies all three conditions of subspaces. Details skipped.