

Section 1.3 Vector Equations

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 - ▶ the **sum** of two vectors u and v is the vector $u + v$ obtained by adding corresponding entries of u and v .
 - ▶ given a vector u and a real number c , the **scalar multiple** of u by c is the vector cu obtained by multiplying each entry in u by c .

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Geometric descriptions of \mathbf{R}^2

- Consider a rectangular coordinate system in the plane. Because each point in the plane is determined by an ordered pair of numbers, we can identify a geometric point (a, b) with the column vector $\begin{bmatrix} a \\ b \end{bmatrix}$.

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- So we may regard \mathbf{R}^2 as the set of all points in the plane.
- If vectors u and v are represented as points in the plane, then the vector $u + v$ corresponds to the fourth vertex of the parallelogram whose other vertices are u , v and 0 .

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- They are represented geometrically by points in a three-dimensional coordinate space, with arrows from the origin.
- If n is a positive integer, \mathbf{R}^n (read r-n) denotes the collection of all lists (or ordered n -tuples) of n real numbers, usually written as $n \times 1$

matrices, such as
$$\begin{bmatrix} u_1 \\ u_2 \\ \dots \\ u_n \end{bmatrix}$$

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 - 8 $1u = u$

Linear combinations of vectors

- Given vectors v_1, v_2, \dots, v_p in \mathbf{R}^n and scalars c_1, c_2, \dots, c_p , the vector y defined by

$$y = c_1 v_1 + c_2 v_2 + \dots + c_p v_p$$

is called a **linear combination** of v_1, \dots, v_p with **weights** c_1, c_2, \dots, c_p .

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- The weights in a linear combination can be any real numbers, including zero.

Linear combinations of vectors—Example

Ex: Let $a_1 = \begin{bmatrix} 1 \\ -2 \\ -5 \end{bmatrix}$, $a_2 = \begin{bmatrix} 2 \\ 5 \\ 6 \end{bmatrix}$ and $b = \begin{bmatrix} 7 \\ 4 \\ -3 \end{bmatrix}$. Determine whether vector b can be written as a linear combination of vectors a_1 and a_2 .

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- This is a **vector equations**.

Solving vector equations—an example

- We can first rewrite the vector equation as a linear systems, by definitions of scalar multiplication and vector addition

$$x_1 \begin{bmatrix} 1 \\ -2 \\ -5 \end{bmatrix} + x_2 \begin{bmatrix} 2 \\ 5 \\ 6 \end{bmatrix} = \begin{bmatrix} 7 \\ 4 \\ -3 \end{bmatrix},$$

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- So we have the following linear system

$$\begin{aligned} x_1 + 2x_2 &= 7 \\ -2x_1 + 5x_2 &= 4 \\ -5x_1 + 6x_2 &= -3 \end{aligned} \quad \text{with augmented matrix} \quad \begin{bmatrix} 1 & 2 & 7 \\ -2 & 5 & 4 \\ -5 & 6 & -3 \end{bmatrix}$$

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- Solve it, we get $x_1 = 3$ and $x_2 = 2$. So $b = 3a_1 + 2a_2$.

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- In general, a vector equation

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has the same solution set as the linear system whose augmented matrix is

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- In particular, the vector b can be generated by a linear combination of vectors a_1, a_2, \dots, a_n **if and only if** there exists a solution to the linear system corresponding to the above augmented matrix.

Span of vectors

- Definition: If v_1, \dots, v_p are vectors in \mathbf{R}^n , then the set of all linear combinations of v_1, \dots, v_p is denoted by $\text{Span}\{v_1, \dots, v_p\}$ and is called the subset of \mathbf{R}^n spanned (or generated) by v_1, \dots, v_p .

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- That is, $Span\{v_1, \dots, v_p\}$ is the collection of all vectors that can be written in the form

$$c_1 v_1 + c_2 v_2 + \dots + c_p v_p,$$

with real numbers (or scalars) c_1, c_2, \dots, c_p .

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- Let u and v be vectors in \mathbf{R}^3 . What is $\text{Span}\{u, v\}$?

1.4 Matrix equations (part 1)

Matrix equation

- Definition: If A is an $m \times n$ matrix, with column vectors a_1, a_2, \dots, a_n , and if x is a vector in \mathbf{R}^n , then the **product of A and x , denoted by Ax** , is the linear combination of the column vectors of A using the corresponding entries in vector x as weights. That is

$$Ax = \begin{bmatrix} a_1 & a_2 & \dots & a_n \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \dots \\ x_n \end{bmatrix} = x_1 a_1 + x_2 a_2 + \dots + x_n a_n.$$

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- Ax is defined only if the number of columns of A equals the number of entries in vector x .

Examples

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- That is,

$$3v_1 - 5v_2 + 7v_3 = [v_1 \quad v_2 \quad v_3] \begin{bmatrix} 3 \\ -5 \\ 7 \end{bmatrix} = Ax.$$

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- As in the previous example, we may write it as matrix equation

$$\begin{bmatrix} 1 & 2 & -1 \\ 0 & -5 & 3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 4 \\ 1 \end{bmatrix}$$

Equivalent formulations

- **Theorem 3:** if A is an $m \times n$ matrix, with column vectors a_1, a_2, \dots, a_n , and if b is a vector in \mathbf{R}^m , then the following three equations have the same solution set

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- So the equation $Ax = b$ has a solution **if and only if** vector b is a linear combination of the column vectors of A .

Existence of solutions—When vectors span \mathbf{R}^m ?

- **Theorem 4:** Let A be an $m \times n$ matrix. Then the following statements are logically equivalent. That is, for a particular A , either they are all true or they are all false.

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 - (c) The column vectors of A span \mathbf{R}^m .
 - (d) The matrix A has a pivot position in every row.

Example

- Ex: Determine if b is in the $\text{Span}\{v_1, v_2, v_3\}$, where vectors v_1, v_2, v_3, b are

$$v_1 = \begin{bmatrix} 1 \\ -4 \\ -3 \end{bmatrix}, v_2 = \begin{bmatrix} 3 \\ 2 \\ -2 \end{bmatrix}, v_3 = \begin{bmatrix} 4 \\ -6 \\ -7 \end{bmatrix}, b = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$$

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- It follows that the system $Ax = b$ is inconsistent for that vector b . So (a) is false.