

Section 4.1 Vector Spaces

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 - ✓ ⑩ $1u = u$. (one)

$$(-1)u = -u$$

$$1 \cdot u = u$$

Examples

Ex1 \mathbf{R}^n is a vector space. (check the ten rules)

$$\begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}$$

$$u + v$$

$$c \cdot u$$

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Define addition by the parallelogram rule, and for each $v \in V$, define cv to be the arrow whose length is $|c|$ times the length of v , pointing in the same direction as v if $c \geq 0$ and otherwise pointing in the opposite direction.



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This gives a vector space.

Ex3 For $n \geq 0$, the set P_n of polynomials of degree at most n . So P_n consists of all polynomials of the form $p(t) = a_0 + a_1t + a_2t^2 + \dots + a_nt^n$, where coefficients a_0, a_1, \dots, a_n and the variable t are real numbers.

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Let $p(t) = a_0 + a_1t + a_2t^2 + \dots + a_nt^n$ and $q(t) = b_0 + b_1t + b_2t^2 + \dots + b_nt^n$, we define addition as

$$(p + q)(t) = (a_0 + b_0) + (a_1 + b_1)t + (a_2 + b_2)t^2 + \dots + (a_n + b_n)t^n$$

and scalar multiplication as

$$(cp)(t) = cp(t) = ca_0 + (ca_1)t + (ca_2)t^2 + \dots + (ca_n)t^n$$

$p(t) = q(t)$ $\Leftrightarrow a_i = b_i$ for all i

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This is a vector space (of polynomials of degree at most n)

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- Conversely, every vector space is a subspace (of itself and possibly of other larger spaces).
- The set consisting of only the zero vector in a vector space V is a subspace of V , called the **zero subspace** and written as $\{0\}$.

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Ex. The vector space \mathbf{R}^2 is NOT a subspace of \mathbf{R}^3 , as \mathbf{R}^2 is not a subset of \mathbf{R}^3 .

$$\begin{bmatrix} 1 \\ 2 \end{bmatrix} \notin \mathbf{R}^3$$



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- Ex. P_n is a subspace of P .
- Ex. The vector space \mathbf{R}^2 is NOT a subspace of \mathbf{R}^3 , as \mathbf{R}^2 is not a subset of \mathbf{R}^3 .
- Ex. The set $H = \{(s, t, 0)^T : s, t \in \mathbf{R}\}$ is a subset of \mathbf{R}^3 . And it is a subspace of \mathbf{R}^3 .

$$\parallel \begin{bmatrix} s \\ t \\ 0 \end{bmatrix}$$

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- Ex. A plane in \mathbf{R}^3 not through the origin is not a subspace of \mathbf{R}^3 .

A subspace spanned by a set

- As the term **linear combination** refers to any sum of scalar multiples of vectors, and $\text{Span}\{v_1, \dots, v_p\}$ denotes the set of all vectors that can be written as linear combinations of v_1, \dots, v_p .

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Proof. we need to verify the ^{four} three conditions:

$$H \subseteq V \quad H \text{ is a subset of } V.$$

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 - ▶ For any $c \in \mathbf{R}$ and $u \in H$, we have $cu = (cs_1)v_1 + (cs_2)v_2 \in H$.
- So H is a subspace of V .

- **Theorem.** If v_1, \dots, v_p are in a vector space V , then $\text{Span}\{v_1, \dots, v_p\}$ is a subspace of V .

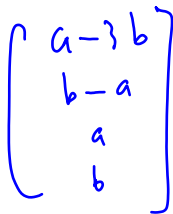
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- Given any subspace H of V , a **spanning (or generating) set** for H is a set $\{v_1, \dots, v_p\}$ in H such that $H = \text{Span}\{v_1, \dots, v_p\}$.

Example

Ex. Let $H = \{(a - 3b, b - a, a, b)^T : a, b \in \mathbf{R}\}$. That is, H is the set of all vectors of the form $(a - 3b, b - a, a, b)^T$ where a and b are arbitrary scalars. Show that H is subspace of \mathbf{R}^4 .

A handwritten blue vector in \mathbf{R}^4 is shown, enclosed in large square brackets. The components are arranged vertically: the first component is $a - 3b$, the second is $b - a$, the third is a , and the fourth is b . A blue arrow originates from the text in the example above and points to the first component of this vector.
$$\begin{bmatrix} a - 3b \\ b - a \\ a \\ b \end{bmatrix}$$

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Proof. The vectors in H can be written as linear combinations:

$$\begin{bmatrix} a - 3b \\ b - a \\ a \\ b \end{bmatrix} = a \begin{bmatrix} 1 \\ -1 \\ 1 \\ 0 \end{bmatrix} + b \begin{bmatrix} -3 \\ 1 \\ 0 \\ 1 \end{bmatrix}$$

$\underbrace{\hspace{10em}}_{v_1} \qquad \underbrace{\hspace{10em}}_{v_2}$

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- So $H = \text{Span}\{v_1, v_2\}$ with $v_1 = (1, -1, 1, 0)^T$ and $v_2 = (-3, 1, 0, 1)^T$. Thus H is a subspace of \mathbf{R}^4 .

Example

- For what values of h will y be in the subspace of \mathbf{R}^3 spanned by v_1, v_2, v_3 if

$$v_1 = \begin{bmatrix} 1 \\ -1 \\ -2 \end{bmatrix}, v_2 = \begin{bmatrix} 5 \\ -4 \\ -7 \end{bmatrix}, v_3 = \begin{bmatrix} -3 \\ 1 \\ 0 \end{bmatrix}, y = \begin{bmatrix} -4 \\ 3 \\ h \end{bmatrix}$$

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$$\begin{bmatrix} 1 & 5 & -3 & -4 \\ -1 & -4 & 1 & 3 \\ -2 & -7 & 0 & h \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 5 & -3 & -4 \\ 0 & 1 & -2 & -1 \\ 0 & 0 & 0 & h-5 \end{bmatrix}$$

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- By looking at its echelon form, we see that the linear system is consistent only if $h - 5 = 0$.

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- For what values of h will y be in the subspace of \mathbf{R}^3 spanned by v_1, v_2, v_3 if

$$v_1 = \begin{bmatrix} 1 \\ -1 \\ -2 \end{bmatrix}, v_2 = \begin{bmatrix} 5 \\ -4 \\ -7 \end{bmatrix}, v_3 = \begin{bmatrix} -3 \\ 1 \\ 0 \end{bmatrix}, y = \begin{bmatrix} -4 \\ 3 \\ h \end{bmatrix}$$

Sol. let $y = x_1 v_1 + x_2 v_2 + x_3 v_3$ with $x_1, x_2, x_3 \in \mathbf{R}$. We then have a linear system whose augmented matrix

$$\begin{bmatrix} 1 & 5 & -3 & -4 \\ -1 & -4 & 1 & 3 \\ -2 & -7 & 0 & h \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 5 & -3 & -4 \\ 0 & 1 & -2 & -1 \\ 0 & 0 & 0 & h-5 \end{bmatrix}$$

- By looking at its echelon form, we see that the linear system is consistent only if $h - 5 = 0$.
- So y is in H if $h = 5$.