

**Important Ideas and Useful Facts:**

- (i) **Abstract vector spaces:** Given a fixed field  $F$ , a *vector space over  $F$*  is an abelian group  $V$  with respect to addition, which is compatible with scalar multiplication by elements of  $F$  (denoted by juxtaposition), in the following respects:

$$(\forall \lambda, \mu \in F)(\forall \mathbf{v}, \mathbf{w} \in V) \quad (\lambda + \mu)\mathbf{v} = \lambda\mathbf{v} + \mu\mathbf{v} \quad \text{and} \quad \lambda(\mathbf{v} + \mathbf{w}) = \lambda\mathbf{v} + \lambda\mathbf{w} ,$$

$$(\forall \lambda, \mu \in F)(\forall \mathbf{v} \in V) \quad \lambda(\mu\mathbf{v}) = (\lambda\mu)\mathbf{v} ,$$

and

$$(\forall \mathbf{v} \in V) \quad 1\mathbf{v} = \mathbf{v} .$$

Here 1 is the multiplicative identity element of  $F$  and the addition symbol  $+$  has to be read in context, belonging either to  $V$  or to  $F$ . It is an important theorem that  $V$  is isomorphic to  $F^n$  for some  $n$  (where  $n$  may be infinite, with an appropriate interpretation).

- (ii) **Vector space isomorphism:** A mapping  $\phi : V \rightarrow W$ , where  $V$  and  $W$  are vector spaces over a field  $F$  is called a *vector space isomorphism* if it is a bijection that preserves addition and scalar multiplication, that is,  $\phi(\mathbf{v} + \mathbf{w}) = \phi(\mathbf{v}) + \phi(\mathbf{w})$  and  $\phi(\lambda\mathbf{v}) = \lambda\phi(\mathbf{v})$  for all  $\mathbf{v}, \mathbf{w} \in V$  and  $\lambda \in F$  (or, equivalently, preserves linear combinations).

- (iii) **Important examples of vector spaces:** Let  $F$  be a field.

- (a) The *trivial vector space* is  $F^0 = \{\mathbf{0}\}$ , consisting of the zero vector with trivial addition and scalar multiplication.
- (b) If  $n \geq 1$  then  $F^n$ , the *Cartesian power*, consisting of all  $n$ -tuples of elements of  $F$ , forms a vector space with respect to coordinate-wise addition and scalar multiplication. We may identify  $n$ -tuples with row vectors of length  $n$ , in which case the vector addition and scalar multiplication of  $n$ -tuples become addition and scalar multiplication of row matrices.
- (c) If  $m, n \geq 1$  then the set  $\text{Mat}_{m,n}$  of all  $m \times n$  matrices forms a vector space with respect to matrix addition and scalar multiplication. In particular,  $\text{Mat}_{1,n}$ , the vector space of row matrices, is identified with  $F^n$ . The vector space  $\text{Mat}_{m,1}$  of column matrices of length  $m$  is isomorphic to  $F^m$  under the mapping that takes a matrix to its transpose.
- (d) If  $n \geq 0$  then the set  $\mathbb{P}_n$  of all polynomials, with coefficients from  $F$ , of degree at most  $n$  forms a vector space with respect to addition of polynomials and multiplication by constants. Then  $\mathbb{P}_n$  is isomorphic to  $F^{n+1}$ .
- (e) Let  $X$  be a nonempty set. Then the set of all functions from  $X$  into  $F$ , denoted by  $F^X$ , forms a vector space with respect to addition of functions and multiplication of a function by a scalar, defined by the following rules, for  $f, g \in F^X$  and  $\lambda \in F$ :

$$(f + g)(x) = f(x) + g(x) \quad \text{and} \quad (\lambda f)(x) = \lambda f(x) \quad \text{for all } x \in X.$$

- (iv) **Subspaces:** A *subspace* of a vector space  $V$  over a field  $F$  is a nonempty subset  $S$  of  $V$  that is closed under vector addition and scalar multiplication, that is, for all  $\mathbf{v}, \mathbf{w} \in S$  and  $\lambda \in F$ ,

$$\mathbf{v} + \mathbf{w} \in S \quad \text{and} \quad \lambda \mathbf{v} \in S ,$$

or, equivalently,  $S$  is closed under taking linear combinations, that is,

$$(\forall \mathbf{v}_1, \mathbf{v}_2 \in S)(\forall \lambda_1, \lambda_2 \in F) \quad \lambda_1 \mathbf{v}_1 + \lambda_2 \mathbf{v}_2 \in S .$$

A subspace  $S$  of a vector space  $V$  becomes a vector space in its own right, using the vector space operations of  $V$  restricted to  $S$ .

- (v) **Intersections of subspaces:** Let  $V$  be a vector space. The intersection of any collection of subspaces of  $V$  is also a subspace of  $V$ . This implies that if  $X$  is any subset of  $V$  then there exists a smallest subspace of  $V$  containing  $X$ , denoted by  $\langle X \rangle$ , and referred to also as the *span of  $X$*  (see more below), namely

$$\langle X \rangle = \bigcap \{S \mid S \text{ is a subspace of } V \text{ containing } X\} ,$$

the intersection of all subspaces of  $V$  containing  $X$ .

- (vi) **Linear combinations:** For  $k \geq 1$ , a *linear combination* of vectors  $\mathbf{v}_1, \dots, \mathbf{v}_k \in V$  is an expression of the form

$$\lambda_1 \mathbf{v}_1 + \lambda_2 \mathbf{v}_2 + \dots + \lambda_k \mathbf{v}_k$$

for some scalars  $\lambda_1, \dots, \lambda_k$ . If  $k = 1$  then this is interpreted as a scalar multiple of  $\mathbf{v}_1$ . Note that since  $0\mathbf{v} = \mathbf{0}$ , for any vector  $\mathbf{v}$ , the zero vector is always a linear combination of any collection of vectors.

- (vii) **The span of a set of vectors:** Let  $X$  be a subset of a vector space  $V$  over a field  $F$ . The *span of  $X$* , denoted by  $\langle X \rangle$  is defined to be  $\{\mathbf{0}\}$ , the trivial subspace of  $V$ , if  $X = \emptyset$ , and otherwise

$$\langle X \rangle = \{\text{all possible linear combinations of finite collections of vectors from } X\} .$$

It follows, in both cases, that  $\langle X \rangle$  is the smallest subspace of  $V$  containing  $X$  (see above). If  $X = \{\mathbf{v}_1, \dots, \mathbf{v}_k\}$  then

$$\langle X \rangle = \{\lambda_1 \mathbf{v}_1 + \lambda_2 \mathbf{v}_2 + \dots + \lambda_k \mathbf{v}_k \mid \lambda_1, \dots, \lambda_k \in F\}$$

- (viii) **Row and column spaces of a matrix:** Let  $M$  be an  $m \times n$  matrix. The *row space* of  $M$  is the vector space of row vectors of length  $n$  spanned by the rows of  $M$ . The *column space* of  $M$  is the vector space of column vectors of length  $m$  spanned by the columns of  $M$ . Two matrices of the same size have the same row [column] space if and only if they are *row [column] equivalent*, that is, can be obtained from one another by elementary row [column] operations. The nonzero rows of any row echelon form for  $M$  span the row space of  $M$  (and in fact form a *basis*, see later). An analogous statement hold for the column space.

- (ix) **Null space of a matrix:** Let  $M$  be an  $m \times n$  matrix over a field  $F$ . The *null space* of  $M$  may refer either to the vector space

$$\{\text{column vectors } \mathbf{v} \text{ of length } n \mid M\mathbf{v} = \mathbf{0}\} ,$$

or the solution space of the associated homogeneous system of  $m$  equations in  $n$  variables:

$$\{\mathbf{v} \in F^n \mid M\mathbf{v}^\top = \mathbf{0}\} .$$

## Tutorial Exercises:

1. Explain how the set of complex numbers

$$\mathbb{C} = \{a + bi \mid a, b \in \mathbb{R}\},$$

where  $i = \sqrt{-1}$ , becomes a vector space over the field  $\mathbb{R}$ . How might one identify complex numbers with geometric vectors in the plane? Find a spanning set for  $\mathbb{C}$  consisting of two elements.

2. Consider the following subsets of the real vector space  $\mathbb{R}^2$ :

$$\begin{aligned} S_1 &= \{(x, y) \mid x + y = 0\}, & S_2 &= \{(x, y) \mid x + y = 1\}, \\ S_3 &= \{(x, y) \mid x + y \geq 0\}, & S_4 &= \{(x, y) \mid x^2 + y^2 = 1\}. \end{aligned}$$

Describe each of these sets geometrically and decide whether it is a subspace of  $\mathbb{R}^2$ .

3. Consider the following subsets of the real vector space  $\mathbb{R}^3$ :

$$\begin{aligned} S_1 &= \{(x, y, z) \mid 2x + 3y + 4z = 0\}, & S_2 &= \{(x, y, z) \mid 2x + 3y + 4z = 1\}, \\ S_3 &= \{(x, y, z) \mid 2x + 3y + 4z \leq 0\}, & S_4 &= \{(x, y, z) \mid x^2 + y^2 + z^2 \leq 1\}. \end{aligned}$$

Describe each of these sets geometrically and decide whether it is a subspace of  $\mathbb{R}^3$ .

4. Working over  $\mathbb{R}$ , determine whether the following matrices have the same or different row spaces:

$$A = \begin{bmatrix} 1 & 1 & 5 \\ 2 & 3 & 13 \end{bmatrix}, \quad B = \begin{bmatrix} 1 & -1 & -2 \\ 3 & -2 & -3 \end{bmatrix}, \quad C = \begin{bmatrix} 1 & -1 & -1 \\ 4 & -3 & -1 \\ 3 & -1 & 3 \end{bmatrix},$$

5. Let  $S_1 = \langle \mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3 \rangle$  be the subspace of  $\mathbb{R}^4$  spanned by the vectors

$$\mathbf{v}_1 = (1, 2, -1, 3), \quad \mathbf{v}_2 = (2, 4, 1, -2), \quad \mathbf{v}_3 = (3, 6, 3, -7),$$

and  $S_2 = \langle \mathbf{w}_1, \mathbf{w}_2 \rangle$  be the subspace of  $\mathbb{R}^4$  spanned by the vectors

$$\mathbf{w}_1 = (1, 2, -4, 11), \quad \mathbf{w}_2 = (2, 4, -5, 14).$$

By row reducing appropriate matrices, verify that  $S_1 = S_2$ .

- 6.\* Let  $V$  be a vector space over a field  $F$ . Prove carefully from the definition of a vector space the following elementary properties:

- (a) The zero vector is unique.
- (b) The negative of a vector is unique.
- (c) For all  $\mathbf{v} \in V$ , we have  $0\mathbf{v} = \mathbf{0}$ , where  $0$  is the zero in  $F$  and  $\mathbf{0}$  is the zero vector.
- (d) For all  $\lambda \in F$ , we have  $\lambda\mathbf{0} = \mathbf{0}$ .
- (e) For all  $\mathbf{v} \in V$ , we have  $(-1)\mathbf{v} = -\mathbf{v}$ , the negative vector.
- (f) For all  $\mathbf{v} \in V$  and  $\lambda \in F$ , we have that  $\lambda\mathbf{v} = \mathbf{0}$  implies  $\lambda = 0$  or  $\mathbf{v} = \mathbf{0}$ .

### Further Exercises:

7. Let  $V$  be a vector space over a field  $F$ . Verify that  $S$  is closed under addition and scalar multiplication if and only if  $\lambda \mathbf{v} + \mu \mathbf{w} \in S$  for all  $\mathbf{v}, \mathbf{w} \in S$  and  $\lambda, \mu \in F$  (that is,  $S$  is closed under taking linear combinations).
8. Let  $V$  be a vector space and suppose that  $S$  and  $T$  are subspaces of  $V$ . Verify that the intersection  $S \cap T$  is a subspace of  $V$ .
9. Explain why a subspace of a vector space is a vector space in its own right, that is, becomes an abelian group with a compatible scalar multiplication.
10. Identify the zero vector and negative vectors in the vector space  $F^X$  of functions from  $X$  to  $F$ , where  $F$  is any field and  $X$  any nonempty set.
11. Let  $m, n \geq 1$  and  $M$  be an  $m \times n$  matrix over a field  $F$ . Verify that the null space of  $M$ , namely

$$S = \{\mathbf{v} \in F^n \mid M\mathbf{v}^\top = \mathbf{0}\},$$

is a subspace of  $F^n$ .

- 12.\* Let  $V$  be any vector space. Verify that every subspace of  $V$  contains the zero vector  $\mathbf{0}$  and that  $\{\mathbf{0}\}$  is a subspace of  $V$ . Deduce that

$$\{\mathbf{0}\} = \bigcap \{S \mid S \text{ is a subspace of } V \text{ containing the empty set}\}.$$

This explains why we define the span of the empty set to be the trivial subspace (and explains why, after we introduce the concepts of *basis* and *dimension*, that the trivial vector space is *zero-dimensional*).

- 13.\* A square matrix  $M$  is *symmetric* if it equals its transpose, that is,  $M = M^\top$ . Verify that, for  $n \geq 1$ , and working over some field  $F$ , the set  $S$  of symmetric  $n \times n$  matrices forms a subspace of the vector space  $\text{Mat}_{n,n}(F)$  of  $n \times n$  matrices over  $F$ . Find a spanning set for  $S$  if  $n = 2$ .
- 14.\* Consider the field  $F = \mathbb{R}$ . Recall that  $\mathbb{P}_n$  denotes the vector space of all real polynomials (which may also be regarded as real polynomial functions) of degree at most  $n$ , where  $n \geq 0$ . Now put

$$\mathbb{P} = \bigcup_{n \geq 0} \mathbb{P}_n.$$

- (a) Verify that  $\mathbb{P}_n$  is a subspace of  $\mathbb{P}$  for each  $n$ , and that  $\mathbb{P}$  is a subspace of  $\mathbb{R}^\mathbb{R}$ .
  - (b) Explain why  $\{1, x, x^2, \dots, x^n\}$  spans  $\mathbb{P}_n$ .
  - (c) Explain why  $\mathbb{P}_n$  and  $\mathbb{R}^{n+1}$  are isomorphic as vector spaces for each  $n \geq 0$ .
  - (d) Explain why no finite subset of  $\mathbb{P}$  can span  $\mathbb{P}$ .
- 15.\* A function  $f : \mathbb{R} \rightarrow \mathbb{R}$  is called *bounded* if there exists some nonnegative real number  $K$  such that  $|f(x)| \leq K$  for all  $x \in \mathbb{R}$ . Prove that the set of all bounded functions is a subspace of the vector space  $\mathbb{R}^\mathbb{R}$  of all real valued real functions.