

Linear Algebra I

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1 Vector Spaces

1.1 Fields

Definition 1.1. A **field** is a set \mathbb{F} on which we have two binary operations: $+$: $\mathbb{F} \times \mathbb{F} \rightarrow \mathbb{F}$ and \cdot : $\mathbb{F} \times \mathbb{F} \rightarrow \mathbb{F}$, satisfying the following axioms:

(F1) (Associativity of addition) For all $a, b, c \in \mathbb{F}$, $a + (b + c) = (a + b) + c$.

(F2) (Commutativity of addition) For all $a, b \in \mathbb{F}$, $a + b = b + a$.

(F3) (Existence of zero) There exists $0 \in \mathbb{F}$ such that, for all $a \in \mathbb{F}$, $a + 0 = a = 0 + a$.

(F4) (Existence of additive inverses) For each $a \in \mathbb{F}$, there exists $-a \in \mathbb{F}$ such that $a + (-a) = 0 = -a + a$.

(F5) (Associativity of multiplication) For all $a, b, c \in \mathbb{F}$, $a(bc) = (ab)c$.

(F6) (Commutativity of multiplication) For all $a, b \in \mathbb{F}$, $ab = ba$.

(F7) (Existence of identity) There exists $1 \in \mathbb{F}$ such that, for all $a \in \mathbb{F}$, $a1 = a = 1a$.

(F8) (Existence of multiplicative inverses) For each $a \in \mathbb{F}$, $a \neq 0$, there exists $a^{-1} \in \mathbb{F}$ such that $aa^{-1} = 1 = a^{-1}a$.

(F9) (Distributivity) For all $a, b, c \in \mathbb{F}$, $a(b + c) = ab + ac$ and $(a + b)c = ac + bc$.

The elements of \mathbb{F} may be called **numbers** or **scalars**, the operations $+$ and \cdot will be called **addition** and **multiplication**, respectively.

Example 1.2.

\mathbb{Q} (the rationals), \mathbb{R} (the reals), and \mathbb{C} (the complex numbers) are fields, while the integers \mathbb{Z} do *not* form a field. Can you see why?

1.2 Finite fields

In this course we will also need finite fields. By this we mean a field \mathbb{F} which is finite as a set. So there are only finitely many numbers in this number system!

Theorem 1.3. *If \mathbb{F} is a finite field then $|\mathbb{F}| = p^f$ for some prime integer p and a positive integer f . Furthermore, for each order p^f , there is exactly one field of that order.*

This theorem tells us how many finite fields exist and what are the possible orders. However, we also need to know what the fields are. We will write \mathbb{F}_q for the only field of order (or size) $q = p^f$.

1.2.1 Prime order

If the order of the field is p (that is, $f = 1$) then the field \mathbb{F}_p is simply \mathbb{Z}_p , the integers modulo p . This means that $\mathbb{F}_p = \{0, 1, \dots, p - 1\}$ and the operations of addition and multiplication on \mathbb{F}_p are the usual addition and multiplication supplemented by subtracting a suitable multiple of p in order to bring the result within the interval $[0, p - 1]$.

Example 1.4. *Suppose we need to work with \mathbb{F}_5 . Then $\mathbb{F}_5 = \{0, 1, 2, 3, 4\}$, and here are examples of addition:*

$$2 + 2 = 4,$$

$$4 + 3 = 7 - 5 = 2.$$

In the first example, the sum, 4 is already in the range, so we do not need to subtract 5. In the second example, we subtract 5 from the sum, 7, in order to bring the result into the range.

Here also are the examples of multiplication:

$$2 \cdot 2 = 4,$$

$$4 \cdot 4 = 16 - 5 - 5 - 5 = 1.$$

In the first example, we are in the range, so we do not subtract any 5s. In the second example, we need to subtract 5 three times in order for the result to be in the target interval.

1.2.2 Non-prime order

In the non-prime case the above method does *not* work! If you need, for example, the field \mathbb{F}_4 of order 4 then you *cannot* simply take \mathbb{Z}_4 , the integers modulo 4. This is not a field! Instead you can take $\mathbb{F}_4 = \{0, 1, a, b\}$, where

addition and multiplication is given by the following tables.

+	0	1	a	b
0	0	1	a	b
1	1	0	b	a
a	a	b	0	1
b	b	a	1	0

·	0	1	a	b
0	0	0	0	0
1	0	1	a	b
a	0	a	b	1
b	0	b	1	a

1.3 Vector Spaces

The following is the main definition of the course.

Definition 1.5. *Suppose \mathbb{F} is a field. A **vector space** over \mathbb{F} is a set V together with two operations: $+$: $V \times V \rightarrow V$ and \cdot : $\mathbb{F} \times V \rightarrow V$, satisfying the following axioms:*

- (VS1) *(Associativity of addition) For all $u, v, w \in V$, $u + (v + w) = (u + v) + w$.*
- (VS2) *(Commutativity of addition) For all $u, v \in V$, $u + v = v + u$.*
- (VS3) *(Existence of zero vector) There exists $0 \in V$ such that $v + 0 = 0 + v = v$ for all $v \in V$.*
- (VS4) *(Existence of additive inverses) For each $u \in V$, there exists $-u \in V$ such that $u + (-u) = 0 = -u + u$.*
- (VS5) *(Associativity of multiplication) For all $a, b \in \mathbb{F}$ and $u \in V$, $a(bu) = (ab)u$.*
- (VS6) *(Distributivity) For all $a, b \in \mathbb{F}$ and $u, v \in V$, $a(u + v) = au + av$ and $(a + b)v = av + bv$.*
- (VS7) *(Identity multiplication) For all $u \in V$, $1u = u$.*

We refer to the elements of V as **vectors**, as opposed to *scalars*, which are the numbers, that is, the elements of \mathbb{F} .

Note that the product of a scalar and a vector is always to be written in this order: first the scalar and then the vector; never the other way around.

The operations $+$ and \cdot will be called **vector addition** and **scalar multiplication**, respectively.

Example 1.6. Let \mathbb{R} be the field of real numbers, $V = \mathbb{R}^2 = \{(x, y) : x, y \in \mathbb{R}\}$.

$(x_1, y_1) = (x_2, y_2)$ if and only if $x_1 = x_2$ and $y_1 = y_2$. We define the vector addition by $(x_1, y_1) + (x_2, y_2) = (x_1 + x_2, y_1 + y_2)$ [coordinate-wise addition] and scalar multiplication by $a(x_1, y_1) = (ax_1, ay_1)$ [coordinate-wise scalar multiplication]. Show that \mathbb{R}^2 is a vector space over \mathbb{R} .

Example 1.7. Let \mathbb{R} be the field of real numbers, $V = \mathbb{R}^2 = \{(x, y) : x, y \in \mathbb{R}\}$.

$(x_1, y_1) = (x_2, y_2)$ if and only if $x_1 = x_2$ and $y_1 = y_2$. We define the vector addition by $(x_1, y_1) + (x_2, y_2) = (x_1 + x_2, y_1 + y_2)$ [coordinate-wise addition] and scalar multiplication by $a(x_1, y_1) = (ax_1, ay_1)$. Test whether \mathbb{R}^2 is a vector space over \mathbb{R} .

Example 1.8. Let \mathbb{F} be any field and $n \in \mathbb{Z}^+$, $\mathbb{F}^n = \{(x_1, \dots, x_n) : x_i \in \mathbb{F} \text{ for } i = 1, \dots, n\}$.

$(x_1, \dots, x_n) = (y_1, \dots, y_n)$ if and only if $x_i = y_i$ for $i = 1, \dots, n$.

We define the vector addition by $(x_1, \dots, x_n) + (y_1, \dots, y_n) = (x_1 + y_1, \dots, x_n + y_n)$ [coordinate-wise addition]

and scalar multiplication by $a(x_1, \dots, x_n) = (ax_1, \dots, ax_n)$ [coordinate-wise scalar multiplication]. Show that \mathbb{F}^n is a vector space over \mathbb{F} .

Example 1.9. Let \mathbb{F} be any field and $m, n \in \mathbb{Z}^+$, $M_{mn}(\mathbb{F}) = \{A : A \text{ is } m - \text{by-} n \text{ matrix in } \mathbb{F}\}$.

Let $+$ be the sum of two matrices and \cdot be the multiplication of a matrix by a constant. Show that $M_{mn}(\mathbb{F})$ is a vector space over \mathbb{F} .

Example 1.10. Let \mathbb{F} be any field and $n \in \mathbb{Z}^+$, $P_n(\mathbb{F}) = \{a_0 + a_1x + \dots + a_nx^n : a_i \in \mathbb{F} \text{ for } i = 0, 1, \dots, n\}$.

$a_0 + a_1x + \dots + a_nx^n = b_0 + b_1x + \dots + b_nx^n$ if and only if $a_i = b_i$ for $i = 0, 1, \dots, n$.

We define the vector addition by $(a_0 + a_1x + \dots + a_nx^n) + (b_0 + b_1x + \dots + b_nx^n) = (a_0 + b_0) + (a_1 + b_1)x + \dots + (a_n + b_n)x^n$ [coordinate wise addition]

and scalar multiplication by $a(a_0 + a_1x + \dots + a_nx^n) = aa_0 + aa_1x + \dots + aa_nx^n$ [coordinate wise scalar multiplication]. Show that $P_n(\mathbb{F})$ is a vector space over \mathbb{F} .

Example 1.11. Let $X \neq \emptyset$ and $V = \{f : f : X \rightarrow \mathbb{R} \text{ is a function}\}$. The vector addition is defined as the sum of two functions $(f+g)(x) = f(x)+g(x)$ for all $x \in X$. The scalar multiplication is defined by $(af)(x) = af(x)$ for all $x \in X$. Show that V is a vector space over \mathbb{R} .

Example 1.12. Let I be any interval subset of \mathbb{R} and $C(I) = \{f : f : I \rightarrow \mathbb{R} \text{ is a continuous function}\}$. Show that $C(I)$ is a vector space over \mathbb{R} .

Example 1.13. Let I be any interval subset of \mathbb{R} , $n \in \mathbb{Z}^+$ and $C^n(I) = \{f : f : I \rightarrow \mathbb{R} \text{ is a differentiable function whose } n\text{th-derivative is continuous}\}$. Show that $C^n(I)$ is a vector space over \mathbb{R} .

Example 1.14 (Subsets of a set). The set 2^Ω of all subsets of a set Ω is a vector space over \mathbb{F}_2 . The addition is provided by the operation of symmetric difference (often denoted by Δ). Recall, that for two subsets A and B of Ω , the symmetric difference $A\Delta B$ of A and B consists of all elements that are in A , but not in B , and all elements that are in B , but not in A . So

$$A\Delta B = (A \setminus B) \cup (B \setminus A) = (A \cup B) \setminus (A \cap B).$$

How do we define multiplication with scalars?

1.4 Basic properties

Here are some basic properties of vector spaces.

Theorem 1.15 (Elementary properties). Suppose V is a vector space over a field \mathbb{F} . Then the following hold:

- (1) (Cancellation in sums) For $u, v, w \in V$, if $u + v = u + w$ then $v = w$.
- (2) For all $u \in V$, $0u = 0$.
- (3) For all $u \in V$, $(-1)u = -u$.
- (4) Also, for all $a \in \mathbb{F}$, $a0 = 0$.

Theorem 1.16. *Suppose V is a vector space over a field \mathbb{F} . Then*

- (1) The zero vector is unique.*
- (2) For each $u \in V$, $-u$ is unique.*

Theorem 1.17. *Suppose V is a vector space over \mathbb{F} . Then:*

- (1) For $a \in \mathbb{F}$ and $u \in V$, if $au = 0$ then either $a = 0$ or $u = 0$.*
- (2) (Cancellation in products)*
 - (a) For $0 \neq a \in \mathbb{F}$ and $u, v \in V$, if $au = av$ then $u = v$.*
 - (b) Also, for $a, b \in \mathbb{F}$ and $0 \neq u \in V$, if $au = bu$ then $a = b$.*

1.5 Subspaces

Definition 1.18. Suppose V is a vector space over \mathbb{F} . A subset $U \subseteq V$ is a **subspace** if it is a vector space over \mathbb{F} in its own right with respect to addition and multiplication inherited from V .

Definition 1.19. Suppose V is a vector space over \mathbb{F} . We say that a subset $U \subseteq V$ is **closed for addition** if, for all $u, v \in U$, $u + v \in U$.

Definition 1.20. Suppose V is a vector space over \mathbb{F} . A subset $U \subseteq V$ is **closed for multiplication with scalars** if, for all $u \in U$ and all $a \in \mathbb{F}$, $au \in U$.

Theorem 1.21 (Subspace criterion I). Suppose V is vector space over \mathbb{F} . A nonempty subset $U \subseteq V$ is a subspace if and only if U is closed for addition and multiplication with scalars. \square

Note that the empty subset $U = \emptyset$ is not a subspace, because (VS3) fails!! There is no zero vector in this U . This is why we have to exclude the empty set in the above theorem.

The two conditions in the theorem can be replaced by a single condition.

Theorem 1.22 (Subspace criterion, II). A non-empty subset U of a vector space V over \mathbb{F} is a subspace of V if and only if, for all $u, v \in U$ and all $a \in \mathbb{F}$, we have that $au + v \in U$.

Example 1.23. In every vector space V , the subset $U = V$ is obviously a subspace. We say that a subspace $U \subseteq V$ is **proper** if $U \neq V$, that is, U is strictly smaller than V .

Hence $U = V$ is the only subspace of V that is not proper.

Similarly, in every vector space V , the subset $\{0\}$ is a subspace. This subspace is called the **trivial** subspace. We will often write 0 for the trivial subspace.

Example 1.24. In \mathbb{R}^2 , the subset $U = \{(x, y) \mid y = 2x\}$ is a subspace.

Example 1.25. In \mathbb{R}^2 , the subset $U = \{(x, y) \mid y = mx; m \in \mathbb{R}\}$ is a subspace.

Example 1.26. In \mathbb{R}^3 , the subset $U = \{(x, y, 0) \mid x, y \in \mathbb{R}\}$ is a subspace.

Example 1.27. In \mathbb{R}^3 , the subset $U = \{(x, y, 1) \mid x, y \in \mathbb{R}\}$ is not a subspace.

Example 1.28 (\mathbb{R}^2 and \mathbb{R}^3). In the plane \mathbb{R}^2 , the subspaces are:

- (1) the trivial subspace containing only the origin 0 ;
- (2) the straight lines through the origin; and
- (3) the improper subspace—the entire plane.

Similarly, in the 3D space \mathbb{R}^3 , we have the following subspaces:

- (1) the trivial subspace;
- (2) the straight lines through the origin;
- (3) the planes through the origin; and
- (4) the improper subspace—the entire space \mathbb{R}^3 .

Example 1.29. In \mathbb{R}^4 , let $U = \{(x, y, z, w) \mid xw = 0\}$. Is U a subspace of \mathbb{R}^4 ? Explain.

Example 1.30. In $M_{2 \times 2}(\mathbb{R})$, the subset $U = \left\{ \begin{bmatrix} a & b \\ c & d \end{bmatrix} \mid a + b = c \right\}$ is a subspace.

1.6 Intersection of subspaces

Theorem 1.31 (Intersection of subspaces). If $\{U_i \mid i \in I\}$ is a collection of subspaces of a vector space V , then the intersection $W := \bigcap_{i \in I} U_i$ is again a subspace of V .

Example 1.32. In $P_2(\mathbb{R})$, let $U_1 = \{a + bx + cx^2 \mid a + 2b - c = 0\}$ and $U_2 = \{a + bx + cx^2 \mid a + b + 2c = 0\}$. Find $U_1 \cap U_2$.

Note that the union of two subspaces need not be a subspace in general. For example, in \mathbb{R}^2 , if $U_1 = \{(x, y) \mid y = x\}$ and $U_2 = \{(x, y) \mid y = -x\}$, then $U_1 \cup U_2$ is not a subspace.

1.7 Generated subspace

Definition 1.33 (Generated subspace). For a vector space V and a subset $X \subseteq V$, the subspace generated by X (denoted $\langle X \rangle$) is the unique minimal subspace of V containing X .

If $X = A \cup B \cup C$, say, we may write $\langle A, B, C \rangle$ instead of $\langle A \cup B \cup C \rangle$. Also, we will write u in place $\{u\}$.

Example 1.34. If the set X consists of a single vector, u , then every subspace containing u also contains all multiples of u . That is, $\langle u \rangle = \{au \mid a \in \mathbb{F}\}$.

For example, if $V = P_3(\mathbb{R})$ and $u = x + 1$ then $\langle u \rangle$ contains polynomials 0 , $x + 1$, $\frac{1}{2}x + \frac{1}{2}$, $-3x - 3$, and in fact, $\langle u \rangle = \{a(x + 1) \mid a \in \mathbb{R}\} = \{ax + a \mid a \in \mathbb{R}\}$.

Example 1.35. Similarly, if $X = \{u, v\}$ then $\langle u, v \rangle = \{au + bv \mid a, b \in \mathbb{F}\}$.

For example, if $V = \mathbb{R}^2$ and $u = (2, 0)$ and $v = (0, 4)$, then $\langle u, v \rangle = \{a(2, 0) + b(0, 4) \mid a, b \in \mathbb{R}\} = \{(2a, 4b) \mid a, b \in \mathbb{R}\}$. For each $(x, y) \in \mathbb{R}^2$, $(x, y) = \frac{x}{2}(2, 0) + \frac{y}{4}(0, 4)$ which implies $(x, y) \in \langle u, v \rangle$. Thus, $\langle u, v \rangle = \mathbb{R}^2$.

Example 1.36. If $\mathbb{F} = \mathbb{F}_2$ and $V = 2^\Omega$, where $\Omega = \{\text{red, blue, yellow, green}\}$, and $u = \{\text{red, yellow}\}$ and $v = \{\text{blue, yellow, green}\}$, then find $\langle u, v \rangle$.

1.8 Sum of subspaces, direct sum

Definition 1.37 (Sum of subspaces). Suppose U and W are two subspaces of a vector space V . Then we define $U + W = \{u + w \mid u \in U, w \in W\}$.

Theorem 1.38. If U and W are subspaces of a vector space V , then $U + W$ is also a subspace of V . \square

Example 1.39. In \mathbb{R}^2 , let $U = \{(x, 0) \mid x \in \mathbb{R}\}$ and $W = \{(0, y) \mid y \in \mathbb{R}\}$. Find $U + W$.

Example 1.40. In \mathbb{R}^3 , let $U = \{(x, y, 0) \mid x, y \in \mathbb{R}\}$ and $W = \{(0, y, z) \mid y, z \in \mathbb{R}\}$. Find $U + W$.

Definition 1.41 (Direct sum). Suppose U and W are subspaces of a vector space V . We say that $D = U + W$ is the **direct sum** of U and W (denoted $D = U \oplus W$) if for every $d \in D$ there exists only one choice of $u \in U$ and $w \in W$ such that $d = u + w$.

Example 1.42. In Example 1.39, \mathbb{R}^2 is the direct sum of U and W . That is, $\mathbb{R}^2 = U \oplus W$. However, in Example 1.40 \mathbb{R}^3 is not the direct sum of U and W .

Theorem 1.43 (Direct sum criterion). *For subspaces U and W of a vector space V , we have that $D = U + W$ is the direct sum of U and W if and only if $U \cap W = 0$.*

Example 1.44. *In \mathbb{R}^2 , let $U = \{(x, y) \mid y = x\}$ and $W = \{(x, y) \mid y = -x\}$. Show that $\mathbb{R}^2 = U \oplus W$.*

1.9 Linear combinations and independence

Definition 1.45. *Let $S = \{u_1, u_2, \dots, u_k\}$ be a set of vectors in a vector space V over a field \mathbb{F} . A vector u in V is called a **linear combination** of the vectors in S if*

$$u = a_1u_1 + a_2u_2 + \dots + a_ku_k$$

for some $a_1, a_2, \dots, a_k \in \mathbb{F}$.

Example 1.46. In \mathbb{R}^2 , test whether the vector $u = (1, 2)$ is a linear combination of the vectors in $S = \{(1, 1), (1, -1)\}$ or not.

Example 1.47. In \mathbb{R}^3 , show that the vector $u = (1, 2, 3)$ is a linear combination of the vectors in $S = \{(1, 1, 1), (-1, 0, 1)\}$.

Definition 1.48. Let $S = \{u_1, u_2, \dots, u_k\}$ be a set of vectors in a vector space V over a field \mathbb{F} . The set S **spans** V , or V is **spanned** by S (denoted $V = \langle S \rangle$), if every vector in V is a linear combination of the vectors in S .

Example 1.49. In \mathbb{R}^2 , let $S = \{(1, 2), (0, -3)\}$. Show that $\mathbb{R}^2 = \langle S \rangle$.

Example 1.50. In \mathbb{R}^3 , let $S = \{(1, 1, 0), (1, 0, 1), (0, 1, 1)\}$. Show that $\mathbb{R}^3 = \langle S \rangle$.

Definition 1.51 (Finite-dimensional spaces). A vector space V over a field \mathbb{F} is called **finite-dimensional** if there exist a finite subset S of V such that $V = \langle S \rangle$. Otherwise, we call the vector space **infinite dimensional**.

Definition 1.52. Let $S = \{u_1, u_2, \dots, u_k\}$ be a set of distinct vectors in a vector space V over a field \mathbb{F} . Then S is said to be **linearly dependent** if there exist scalars $a_1, a_2, \dots, a_k \in \mathbb{F}$ not all zero, such that

$$a_1u_1 + a_2u_2 + \dots + a_ku_k = 0.$$

Otherwise, S is called **linearly independent**. That is, S is linearly independent if $a_1u_1 + a_2u_2 + \dots + a_ku_k = 0$ holds only when

$$a_1 = a_2 = \dots = a_k = 0.$$

Example 1.53. In \mathbb{R}^3 , let $S = \{(1, 2, 0), (-1, 0, 1), (0, 1, 1)\}$. Show that S is linearly independent.

Example 1.54. In $P_2(\mathbb{R})$, let $S = \{2 + x + x^2, x + 2x^2, 2 + 2x + 3x^2\}$. Test whether S is linearly dependent or linearly independent.

Theorem 1.55. Let $S = \{u_1, u_2, \dots, u_n\}$ be a set of nonzero vectors in a vector space V over a field \mathbb{F} . Then S is linearly dependent if and only if one of the vectors u_j is a linear combination of the preceding vectors in S .

Example 1.56. In \mathbb{R}^3 , let $S = \{(1, 2, -1), (1, -2, 1), (-3, 2, -1), (2, 0, 0)\}$. Test whether S is linearly dependent or not.

1.10 Bases

Definition 1.57 (Basis). *A set of vectors $S = \{u_1, u_2, \dots, u_k\}$ in a vector space V over a field \mathbb{F} is called a **basis** for V if S spans V (that is $V = \langle S \rangle$) and S is linearly independent.*

Theorem 1.58 (Existence of bases). *Every vector space has a basis.*

Example 1.59. *In \mathbb{R}^2 , let $S = \{(1, 0), (0, 1)\}$. Then S is a basis for \mathbb{R}^2 , called the **standard (natural) basis** for \mathbb{R}^2 .*

Example 1.60. *In \mathbb{R}^3 , let $S = \{(1, 0, 0), (0, 1, 0), (0, 0, 1)\}$. Then S is a basis for \mathbb{R}^3 , called the **standard (natural) basis** for \mathbb{R}^3 .*

Example 1.61. *In $P_2(\mathbb{R})$, let $S = \{1, x, x^2\}$. Then S is a basis for $P_2(\mathbb{R})$, called the **standard (natural) basis** for $P_2(\mathbb{R})$.*

Example 1.62. *The set $S = \left\{ \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} \right\}$ is a basis for $M_{2 \times 2}(\mathbb{R})$, called the **standard (natural) basis** for $M_{2 \times 2}(\mathbb{R})$.*

Example 1.63. *Show that the set $U = \{1 + x^2, -1 + x, 2 + 2x\}$ is a basis for $P_2(\mathbb{R})$.*

Theorem 1.64. *If $S = \{u_1, u_2, \dots, u_n\}$ is a basis for a vector space V , then every vector in V can be written in one and only one way as a linear combination of the vectors in S .*

Theorem 1.65. *If $S = \{u_1, u_2, \dots, u_n\}$ is a set of nonzero vectors spanning a vector space V , then S contains a basis T for V .*

Theorem 1.66. *If $S = \{u_1, u_2, \dots, u_n\}$ is a basis for a vector space V and $T = \{v_1, v_2, \dots, v_r\}$ is a linearly independent set of vectors in V , then $n \geq r$.*

Corollary 1.67. *If $S = \{u_1, u_2, \dots, u_n\}$ and $T = \{v_1, v_2, \dots, v_m\}$ are bases for a vector space V , then $n = m$.*

Definition 1.68. *The **dimension** of a nonzero vector space V over a field \mathbb{F} is the number of vectors in a basis for V . We often write **dim** V (or **dim** $_{\mathbb{F}}V$) for the dimension of V . We also define the dimension of $\{0\}$ to be zero.*

Example 1.69.

1. $\dim \mathbb{R}^2 = 2$.
2. $\dim \mathbb{R}^n = n$.
3. If \mathbb{F} is any field, then $\dim_{\mathbb{F}} \mathbb{F}^n = n$.
4. $\dim_{\mathbb{R}} \mathbb{C}^2 = 4$.
5. $\dim P_n(\mathbb{R}) = n + 1$.

Theorem 1.70. *If S is a linearly independent set of vectors in a finite-dimensional vector space V , then there is a basis T for V which contains S .*

Example 1.71. *Find a basis for \mathbb{R}^3 containing the vectors $(1, 0, 1)$ and $(0, -1, 1)$.*

Theorem 1.72. *Let V be an n -dimensional vector space.*

1. *If $S = \{u_1, u_2, \dots, u_n\}$ is a linearly independent set of vectors in V , then S is a basis for V .*
2. *If $S = \{u_1, u_2, \dots, u_n\}$ spans V , then S is a basis for V .*

Example 1.73. *Show that the set $S = \{(1, 1, 0), (1, 0, 1), (0, 1, 1)\}$ is a basis for \mathbb{R}^3 .*

Theorem 1.74. *Suppose U is a subspace of a finite-dimensional vector space V . Then*

1. $\dim U \leq \dim V$.
2. *If $\dim U = \dim V$ then $U = V$.*

1.11 Dimension of the sum of subspaces

Theorem 1.75 (Dimension of sum). *Suppose V is a vector space and suppose U and W are finite-dimensional subspaces of V . Then*

$$\dim(U + W) = \dim U + \dim W - \dim(U \cap W).$$

Example 1.76. *In \mathbb{R}^3 , let $U = \{(x, y, z) \mid z = x + 2y\}$ and $W = \{(x, y, z) \mid x = -y\}$. Find $\dim(U + W)$ and $U + W$.*

Theorem 1.77. *Suppose U and W are finite-dimensional subspaces of a vector space V . Let $S = U + W$. Then $S = U \oplus W$ if and only if $\dim S = \dim U + \dim W$. □*

1.12 Coordinates

Definition 1.78. Let $S = \{u_1, u_2, \dots, u_n\}$ be a basis for a vector space V over a field \mathbb{F} . Then every vector u in V can be written as $u = a_1u_1 + a_2u_2 + \dots + a_nu_n$, where $a_1, a_2, \dots, a_n \in \mathbb{F}$. The vector $X = (a_1, a_2, \dots, a_n)$ is called a **coordinate vector** of u relative to the basis S .

Example 1.79. In $P_2(\mathbb{R})$, find the coordinate vector of $u = 3 - x^2$ relative to the basis $S = \{3, -1 + x, x^2\}$.

Definition 1.80. Let $S = \{u_1, u_2, \dots, u_n\}$ and $S^* = \{u_1^*, u_2^*, \dots, u_n^*\}$ be bases for a vector space V over a field \mathbb{F} . Then every vector $u_i \in S$ can be written as a linear combination of the vectors in S^* . That is,

$$u_1 = a_{11}u_1^* + a_{12}u_2^* + \dots + a_{1n}u_n^*,$$

$$u_2 = a_{21}u_1^* + a_{22}u_2^* + \dots + a_{2n}u_n^*,$$

\vdots

$$u_n = a_{n1}u_1^* + a_{n2}u_2^* + \dots + a_{nn}u_n^*. \text{ Let } P = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & & \ddots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{bmatrix}. \text{ Then } P$$

is called a **transition matrix** from the basis S to the basis S^* .

Example 1.81. Find the transition matrix from the basis $S = \{(2, 3), (0, 3)\}$ for \mathbb{R}^2 to the basis $S^* = \{(-1, 0), (3, 3)\}$.

Example 1.82. Let $P = \begin{bmatrix} \frac{1}{\sqrt{5}} & \frac{2}{\sqrt{5}} \\ -\frac{2}{\sqrt{5}} & \frac{1}{\sqrt{5}} \end{bmatrix}$ be a transition matrix from the basis $S = \{(2, 1), (0, 3)\}$ for \mathbb{R}^2 to the basis $S^* = \{u_1^*, u_2^*\}$. Find u_1^* and u_2^* .

Theorem 1.83. *Let S and S^* be two bases for a f.d.v.s V over a field \mathbb{F} . Let P be a transition matrix from S to S^* . If X is a coordinate vector of $u \in V$ with respect to S , then $X^* = XP$ is a coordinate vector of u with respect to S^* .*

Theorem 1.84. *Let S and S^* be bases for a vector space V . If there exist a matrix $P = (p_{ij})$ such that for any $u \in V$, the vector coordinate of u is X and X^* relative to S and S^* , respectively, and $X^* = XP$, then P is a transition matrix from S to S^* .*

Theorem 1.85. *Let S, S^* and S^{**} be bases for a vector space V . Let P be a transition matrix from S to S^* and Q be a transition matrix from S^* to S^{**} . Then PQ is a transition matrix from S to S^{**} .*

Theorem 1.86. *Let S and S^* be bases for a vector space V . If P is a transition matrix from S to S^* , then P^{-1} is a transition matrix from S^* to S .*