

5.16 Modern Algebra

3. Let V be the set of all polynomials in $F[x]$ of degree $\leq n$. Let $S = \{1, x, x^2, \dots, x^n\}$. Then $L(S) = V$ and hence V is finite-dimensional.
4. C is a finite-dimensional vector space over \mathbf{R} , since $L(\{1, i\}) = C$.
5. In $M_2(\mathbf{R})$ consider the set S consisting of the matrices

$$A = \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}; B = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix};$$

$$C = \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}; D = \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}. \text{ Then}$$

$$\begin{bmatrix} a & b \\ c & d \end{bmatrix} = aA + bB + cC + dD.$$

Hence $L(S) = M_2(\mathbf{R})$ so that $M_2(\mathbf{R})$ is finite-dimensional.

Note. All the vector spaces we have considered above are finite dimensional. However there are vector spaces which cannot be spanned by a finite number of vectors. For example, consider $\mathbf{R}[x]$. Let S be any finite subset of $\mathbf{R}[x]$. Let f be a polynomial of maximum degree in S . Let $\deg f = n$. Then any element of $L(S)$ is a polynomial of degree $\leq n$ and hence $L(S) \neq \mathbf{R}[x]$. Thus $\mathbf{R}[x]$ is not finite-dimensional.

Throughout the rest of this chapter all the vector spaces we consider are finite dimensional.

Although we have defined what is meant by a finite dimensional space we have not yet defined what is meant by the *dimension* of a vector space. We now proceed to introduce the concepts necessary to define the dimension of a finite dimensional vector space.

Consider the vectors $e_1 = (1, 0, 0)$,

$e_2 = (0, 1, 0)$, $e_3 = (0, 0, 1)$ in $V_3(\mathbf{R})$.

Suppose that $\alpha_1 e_1 + \alpha_2 e_2 + \alpha_3 e_3 = 0$.

Then $(\alpha_1, 0, 0) + (0, \alpha_2, 0) + (0, 0, \alpha_3) = (0, 0, 0)$.

$$\therefore (\alpha_1, \alpha_2, \alpha_3) = (0, 0, 0).$$

$$\therefore \alpha_1 = \alpha_2 = \alpha_3 = 0$$

(i.e) $\alpha_1 e_1 + \alpha_2 e_2 + \alpha_3 e_3 = 0$ iff $\alpha_1 = \alpha_2 = \alpha_3 = 0$.

Thus a linear combination of the vectors e_1, e_2 and e_3 will yield the zero vector iff all the coefficients are zero.

Definition. Let V be a vector space over a field F . A finite set of vectors v_1, v_2, \dots, v_n in V is said to be *linearly independent* if

$$\alpha_1 v_1 + \alpha_2 v_2 + \dots + \alpha_n v_n = 0 \\ \Rightarrow \alpha_1 = \alpha_2 = \dots = \alpha_n = 0.$$

If v_1, v_2, \dots, v_n are not linearly independent, then they are said to be *linearly dependent*.

Note. If v_1, v_2, \dots, v_n are linearly dependent, then there exist scalars $\alpha_1, \alpha_2, \dots, \alpha_n$ not all zero, such that $\alpha_1 v_1 + \dots + \alpha_n v_n = 0$.

Examples

1. In $V_n(F)$, $\{e_1, e_2, \dots, e_n\}$ is a linearly independent set of vectors, for,

$$\alpha_1 e_1 + \alpha_2 e_2 + \dots + \alpha_n e_n = 0.$$

$$\Rightarrow \alpha_1(1, 0, \dots, 0) + \alpha_2(0, 1, \dots, 0)$$

$$+ \dots + \alpha_n(0, 0, \dots, 1) \\ = (0, 0, \dots, 0)$$

$$\Rightarrow (\alpha_1, \alpha_2, \dots, \alpha_n) = (0, 0, \dots, 0)$$

$$\Rightarrow \alpha_1 = \alpha_2 = \dots = \alpha_n = 0.$$

2. In $V_3(\mathbf{R})$ the vectors $(1, 2, 1)$, $(2, 1, 0)$ and $(1, -1, 2)$ are linearly independent. For, let

$$\alpha_1(1, 2, 1) + \alpha_2(2, 1, 0)$$

$$+ \alpha_3(1, -1, 2) = (0, 0, 0).$$

$$\therefore (\alpha_1 + 2\alpha_2 + \alpha_3, 2\alpha_1 + \alpha_2 - \alpha_3, \alpha_1 + 2\alpha_3) = (0, 0, 0).$$

$$\therefore \alpha_1 + 2\alpha_2 + \alpha_3 = 0 \dots \dots (1)$$

$$2\alpha_1 + \alpha_2 - \alpha_3 = 0 \dots \dots (2)$$

$$\alpha_1 + 2\alpha_3 = 0 \dots \dots (3)$$

Solving equations (1), (2) and (3) we get

$$\alpha_1 = \alpha_2 = \alpha_3 = 0.$$

\therefore The given vectors are linearly independent.

3. In $V_3(\mathbb{R})$ the vectors $(1, 4, -2)$, $(-2, 1, 3)$ and $(-4, 11, 5)$ are linearly dependent. For let

$$\alpha_1(1, 4, -2) + \alpha_2(-2, 1, 3) + \alpha_3(-4, 11, 5) = (0, 0, 0)$$

$$\therefore \alpha_1 - 2\alpha_2 - 4\alpha_3 = 0 \dots \dots (1)$$

$$4\alpha_1 + \alpha_2 + 11\alpha_3 = 0 \dots \dots (2)$$

$$-2\alpha_1 + 3\alpha_2 + 5\alpha_3 = 0 \dots \dots (3)$$

From (1) and (2),

$$\frac{\alpha_1}{-18} = \frac{\alpha_2}{-27} = \frac{\alpha_3}{9} = k \text{ (say)}$$

$$\therefore \alpha_1 = -18k, \alpha_2 = -27k, \alpha_3 = 9k.$$

These values of α_1, α_2 and α_3 , for any k satisfy (3) also.

Taking $k = 1$ we get

$$\alpha_1 = -18, \alpha_2 = -27, \alpha_3 = 9 \text{ as a non-trivial solution.}$$

Hence the three vectors are linearly dependent.

4. Let V be a vector space over a field F . Then any subset S of V containing the zero vector is linearly dependent.

Proof. Let $S = \{0, v_1, \dots, v_n\}$

Clearly $\alpha 0 + 0v_1 + 0v_2 + \dots + 0v_n = 0$ where α is any element of F . Hence for any $\alpha \neq 0$, we get a non-trivial linear combination of vectors in S giving the zero vector. Hence S is linearly dependent.

Exercises

- Determine whether the following sets of vector are linearly independent or linearly dependent in $V_3(\mathbb{R})$.
 - $\{(1, 0, 0), (0, 1, 0), (1, 1, 0)\}$.
 - $\{(1, 2, 3), (2, 3, 1)\}$.
 - $\{(1, 2, 3), (4, 1, 5), (-4, 6, 2)\}$.
 - $\{(0, 0, 0), (2, 5, 3), (-1, 0, 6)\}$.
 - $\{(1, 0, 0), (1, 1, 0), (1, 1, 1), (0, 1, 0)\}$.
- Determine whether the following sets of vectors are linearly independent or not.
 - $\{(1, 1, 0, 0), (0, 0, 1, 1), (1, 0, 0, 4), (0, 0, 0, 2)\}$ in $V_4(\mathbb{R})$.

- $\{(2i, 1, 0), (2, -i, 1), (0, 1, +i, -i)\}$ in $V_3(\mathbb{C})$.
- $\{(\pi, 0, 0), (0, e, 0), (0, 0, \sqrt{5})\}$ in $V_3(\mathbb{R})$.
- $V =$ the set of all polynomials of degree $\leq n$ in $\mathbb{R}[x]$ and

$$S = \{1, x, x^2, \dots, x^n\}.$$

- $\left\{ \begin{bmatrix} 2 & 0 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 2 & 0 \end{bmatrix}, \begin{bmatrix} 0 & -1 \\ 3 & 0 \end{bmatrix} \right\}$ in $M_2(\mathbb{R})$.

- In $V_3(\mathbb{Z}_5)$ determine whether the following sets of vectors are linearly dependent.
 - $\{(1, 3, 2), (2, 1, 3)\}$
 - $\{(1, 1, 2), (2, 1, 0), (0, 4, 1)\}$.
- In $V_2(\mathbb{R})$ prove that the vectors (a, b) and (c, d) are linearly dependent iff $ad - bc = 0$.
- Let $\{v_1, v_2, v_3\}$ be a linearly independent set of vectors in $V_3(\mathbb{R})$. Show that
 - $\{v_1 + v_2, v_2 + v_3, v_3 + v_1\}$ is linearly independent.
 - $\{2v_1 + v_2, v_1 + v_2, v_1 - v_3\}$ is linearly independent.
- If the vectors $(0, 1, a)$, $(1, a, 1)$ and $(a, 1, 0)$ of $V_3(\mathbb{R})$ are linearly dependent then find the value of a .

Answers.

- (b) is linearly independent.
- (a), (b), (c), (d) and (e) are linearly independent.
- (a) is linearly independent 6. $a = 0, \pm\sqrt{2}$.

Theorem 5.11. Any subset of a linearly independent set is linearly independent.

Proof. Let V be a vector space over a field F .

Let $S = \{v_1, v_2, \dots, v_n\}$ be a linearly independent set.

Let S' be a subset of S . Without loss of generality we take $S' = \{v_1, v_2, \dots, v_k\}$ where $k \leq n$.

Suppose S' is a linearly dependent set. Then there exist $\alpha_1, \alpha_2, \dots, \alpha_k$ in F not all zero, such that

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$$\alpha_1 v_1 + \alpha_2 v_2 + \dots + \alpha_k v_k = 0.$$

Hence $\alpha_1 v_1 + \alpha_2 v_2 + \dots + \alpha_k v_k + 0v_{k+1} + \dots + 0v_n = 0$ is a non-trivial linear combination giving the zero vector.

Here S is a linearly dependent set which is a contradiction.

Hence S' is linearly independent.

Theorem 5.12. Any set containing a linearly dependent set is also linearly dependent.

Proof. Let V be a vector space. Let S be a linearly dependent set. Let $S' \supset S$.

If S' is linearly independent S is also linearly independent (by theorem 5.11) which is a contradiction. Hence S' is linearly dependent.

Theorem 5.13. Let $S = \{v_1, v_2, \dots, v_n\}$ be a linearly independent set of vectors in a vector space V over a field F . Then every element of $L(S)$ can be uniquely written in the form

$$\alpha_1 v_1 + \alpha_2 v_2 + \dots + \alpha_n v_n, \text{ where } \alpha_i \in F.$$

Proof. By definition every element of $L(S)$ is of the form

$$\alpha_1 v_1 + \alpha_2 v_2 + \dots + \alpha_n v_n.$$

Now, let $\alpha_1 v_1 + \alpha_2 v_2 + \dots + \alpha_n v_n = \beta_1 v_1 + \beta_2 v_2 + \dots + \alpha_n v_n$.

$$\text{Hence } (\alpha_i - \beta_1)v_1 + (\alpha_2 - \beta_2)v_2 + \dots +$$

$$(\alpha_n - \beta_n)v_n = 0.$$

Since S is a linearly independent set, $\alpha_i - \beta_i = 0$ for all i .

$\therefore \alpha_i = \beta_i$ for all i . Hence the theorem.

Theorem 5.14. $S = \{v_1, v_2, \dots, v_n\}$ is a linearly dependent set of vectors in V iff there exists a vector $v_k \in S$ such that v_k is a linear combination of the preceding vectors v_1, v_2, \dots, v_{k-1} .

Proof. Suppose v_1, v_2, \dots, v_n are linearly dependent.

Then there exist $\alpha_1, \alpha_2, \dots, \alpha_n \in F$, not all zero, such that $\alpha_1 v_1 + \alpha_2 v_2 + \dots + \alpha_n v_n = 0$.

Let k be the largest integer for which $\alpha_k \neq 0$.

Then $\alpha_1 v_1 + \dots + \alpha_k v_k = 0$.

$$\therefore \alpha_k v_k = -\alpha_1 v_1 - \alpha_2 v_2 - \dots - \alpha_{k-1} v_{k-1}.$$

$$\therefore v_k = (-\alpha_k^{-1} \alpha_1) v_1 + \dots + (-\alpha_k^{-1} \alpha_{k-1}) v_{k-1}.$$

$\therefore v_k$ is a linear combination of the preceding vectors.

Conversely, suppose there exists a vector v_k such that $v_k = \alpha_1 v_1 + \dots + \alpha_{k-1} v_{k-1}$.

$$\text{Hence } -\alpha_1 v_1 - \dots - \alpha_{k-1} v_{k-1} + v_k + 0v_{k+1} + \dots + 0v_n = 0.$$

Since the coefficient of $v_k = 1$, we have

$S = \{v_1, \dots, v_n\}$ is linearly dependent.

Example.

In $V_3(\mathbf{R})$, let $S = \{(1, 0, 0), (0, 1, 0), (0, 0, 1), (1, 1, 1)\}$

$$\text{Here } (1, 1, 1) = (1, 0, 0) + (0, 1, 0) + (0, 0, 1).$$

Thus $(1, 1, 1)$ is a linear combination of the preceding vectors. Hence S is a linearly dependent set.

Theorem 5.15. Let V be a vector space over F . Let $S = \{v_1, v_2, \dots, v_n\}$ and $L(S) = W$. Then there exists a linearly independent subset S' of S such that $L(S') = W$.

Proof. Let $S = \{v_1, v_2, \dots, v_n\}$.

If S is linearly independent there is nothing to prove.

If not, let v_k be the first vector in S which is a linear combination of the preceding vectors.

Let $S_1 = \{v_1, v_2, \dots, v_{k-1}, v_{k+1}, \dots, v_n\}$.

(ie) S_1 is obtained by deleting the vector v_k from S .

We claim that $L(S_1) = L(S) = W$.

Since $S_1 \subseteq S$, $L(S_1) \subseteq L(S)$. (refer theorem 5.10).

Not, let $v \in L(S)$.

$$\text{Then } v = \alpha_1 v_1 + \dots + \alpha_k v_k + \dots + \alpha_n v_n.$$

Now, v_k is a linear combination of the preceding vectors.

$$\text{Let } v_k = \beta_1 v_1 + \dots + \beta_{k-1} v_{k-1}.$$

$$\text{Hence } v = \alpha_1 v_1 + \dots + \alpha_{k-1} v_{k-1} + \alpha_k (\beta_1 v_1 + \dots + \beta_{k-1} v_{k-1}) + \alpha_{k+1} v_{k+1} + \dots + \alpha_n v_n.$$

$\therefore v$ can be expressed as a linear combination of the vectors of S_1 so that $v \in L(S_1)$. Hence $L(S) \subseteq L(S_1)$

Thus $L(S) = L(S_1) = W$.

Now, if S_1 is linearly independent, the proof is complete.

If not, we continue the above process of removing a vector from S_1 , which is a linear combination of the preceding vectors until we arrive at a linearly independent subset S' of S such that $L(S') = W$.

5.6. Basis and Dimension

Definition. A linearly independent subset S of a vector space V which spans the whole space V is called a *basis* of the vector space.

Theorem 5.16. Any finite-dimensional vector space V contains a finite number of linearly independent vectors which span V . (ie) A finite dimensional vector space has a basis consisting of a finite number of vectors.

Proof. Since V is finite dimensional there exists a finite subset S of V such that $L(S) = V$. By theorem 5.15 this set S contains a linearly independent subset $S' = \{v_1, v_2, \dots, v_n\}$ such that

$$L(S') = L(S) = V.$$

Hence S' is a basis for V .

Theorem 5.17. Let V be a vector space over a field F . Then $S = \{v_1, v_2, \dots, v_n\}$ is a basis for V iff every element of V can be uniquely expressed as a linear combination of element of S .

Proof. Let S be a basis for V .

Then by definition S is linearly independent and $L(S) = V$. Hence by theorem 5.13 every element of V can be uniquely expressed as a linear combination of elements of S .

Conversely, suppose every element of V can be uniquely expressed as a linear combination of elements of S .

Clearly $L(S) = V$.

Now, let $\alpha_1 v_1 + \alpha_2 v_2 + \dots + \alpha_n v_n = 0$.

Also, $0v_1 + 0v_2 + \dots + 0v_n = 0$.

Thus we have expressed 0 as a linear combination of vectors of S in two ways.

\therefore By hypothesis $\alpha_1 = \alpha_2 = \dots = \alpha_n = 0$.

Hence S is linearly independent. Hence S is a basis.

Examples

- $S = \{(1, 0, 0), (0, 1, 0), (0, 0, 1)\}$ is a basis for $V_3(\mathbf{R})$ for, $(a, b, c) = a(1, 0, 0) +$

$$b(0, 1, 0) + c(0, 0, 1).$$

\therefore Any vector (a, b, c) of $V_3(\mathbf{R})$ has been expressed uniquely as a linear combination of the elements of S and hence S is a basis for $V_3(\mathbf{R})$

- $S = \{e_1, e_2, \dots, e_n\}$ is a basis for $V_n(F)$. This is known as the *standard basis* for $V_n(F)$.

- $S = \{(1, 0, 0), (0, 1, 0), (1, 1, 1)\}$ is a basis for $V_3(\mathbf{R})$.

Proof. We shall show that any element (a, b, c) of $V_3(\mathbf{R})$ can be uniquely expressed as a linear combination of the vectors of S .

$$\text{Let } (a, b, c) = \alpha(1, 0, 0) + \beta(0, 1, 0) + \gamma(1, 1, 1)$$

Then $\alpha + \gamma = a, \beta + \gamma = b, \gamma = c$.

Hence $\alpha = a - c$ and $\beta = b - c$.

$$\text{Thus } (a, b, c) = (a - c)(1, 0, 0) + (b - c)(0, 1, 0) + c(1, 1, 1).$$

$\therefore S$ is a basis for $V_3(\mathbf{R})$.

- $S = \{1\}$ is a basis for the vector space \mathbf{R} over \mathbf{R} .

- $S = \left\{ \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix} \right\}$ is a basis for $M_2(\mathbf{R})$, since any matrix $\begin{bmatrix} a & b \\ c & d \end{bmatrix}$ can be uniquely written as

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix} = a \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} + b \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} + c \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix} + d \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix}$$

6. $\{1, i\}$ is a basis for the vector space \mathbf{C} over \mathbf{R} .
7. Let V be the set of all polynomials of degree $\leq n$ in $\mathbf{R}[x]$. Then $\{1, x, x^2, \dots, x^n\}$ is a basis for V .
8. $\{(1, 0), (i, 0), (0, 1), (0, i)\}$ is a basis, for the vector space $\mathbf{C} \times \mathbf{C}$ over \mathbf{R} , for $(a + ib, c + id) = a(1, 0) + b(i, 0) + c(0, 1) + d(0, i)$.
9. $S = \{(1, 0, 0), (0, 1, 0), (1, 1, 1), (1, 1, 0)\}$ spans the vector space $V_3(\mathbf{R})$ but is not a basis.

Proof. Let $S' = \{(1, 0, 0), (0, 1, 0), (1, 1, 1)\}$.

Then $L(S') = V_3(\mathbf{R})$ (refer example 3).

Now, since $S \subseteq S'$, we get $L(S) = V_3(\mathbf{R})$.

Thus S spans $V_3(\mathbf{R})$.

But S is linearly dependent since

$$(1, 1, 0) = (1, 0, 0) + (0, 1, 0).$$

Hence S is not a basis.

10. $S = \{(1, 0, 0), (1, 1, 0)\}$ is linearly independent but not a basis of $V_3(\mathbf{R})$.

Proof. Let $\alpha(1, 0, 0) + \beta(1, 1, 0) = (0, 0, 0)$. Then $\alpha + \beta = 0$ and $\beta = 0$.

$\therefore \alpha = \beta = 0$. Hence S is linearly independent.

Also $L(S) = \{(a, b, 0) / a, b \in \mathbf{R}\} \neq V_3(\mathbf{R})$.

$\therefore S$ is not a basis.

Exercises

1. Show that the following three vectors from a basis for $V_3(\mathbf{R})$.
- (a) $(1, 2, -3), (2, 5, 1), (-1, 1, 4)$
- (b) $(1, 1, 0), (0, 1, 1), (1, 0, 1)$
- (c) $(2, -3, 1), (0, 1, 2), (1, 1, 2)$
2. Show that the following sets of vectors do not form a basis for $V_3(\mathbf{R})$.
- (a) $\{(1, 0, 0), (1, 1, 0)\}$
- (b) $\{(1, 2, 1), (1, 3, 5), (-1, 0, 1), (1, -1, 2)\}$

(c) $\{(0, 0, 0), (1, 0, 0), (0, 1, 0), (0, 0, 1)\}$

(d) $\{(3, 2, 1), (3, 1, 5), (3, 4, -7)\}$

(e) $\{(1, 2, 3), (2, 3, 4), (3, 4, 5)\}$

3. Show that $(1, i, 0), (2i, 1, 1), (0, 1 + i, 1 - i)$ form a basis for $V_3(\mathbf{C})$.

4. Find a basis for the vector space consisting of all matrices of the form

(a) $\begin{pmatrix} a & b \\ 0 & d \end{pmatrix}$ (b) $\begin{pmatrix} a & 0 \\ 0 & b \end{pmatrix}$

5. If $\{v_1, v_2, v_3\}$ is a basis for $V_3(\mathbf{R})$, show that $\{v_1 + v_2, v_2 + v_3, v_3 + v_1\}$ is also a basis. Is this true in (a) $V_3(\mathbf{Z}_2)$ (b) $V_3(\mathbf{Z}_3)$?

Answers.

4. (a) $\left\{ \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} \right\}$

(b) $\left\{ \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} \right\}$

Theorem 5.18. Let V be a vector space over a field F . Let $S = \{v_1, v_2, \dots, v_n\}$ span V . Let

$S = \{w_1, w_2, \dots, w_m\}$ be a linearly independent set of vectors in V . Then $m \leq n$.

Proof. Since $L(S) = V$, every vector in V and in particular w_1 , is a linear combination of v_1, v_2, \dots, v_n .

Hence $S_1 = \{w_1, v_1, v_2, \dots, v_n\}$ is a linearly dependent set of vectors. Hence there exists a vector $v_k \neq w_1$ in S_1 which is a linear combination of the preceding vectors.

Let $S_2 = \{w_1, v_1, \dots, v_{k-1}, v_{k+1}, \dots, v_n\}$.

Clearly, $L(S_2) = V$.

Hence w_2 is a linear combination of the vectors in S_2 .

Hence $S_3 = \{w_2, w_1, v_1, \dots, v_{k-1}, v_{k+1}, \dots, v_n\}$ is linearly dependent. Hence there exists a vector in S_3 which is a linear combination of the preceding vectors. Since the w_i 's are linearly independent, this vector cannot be w_2 or w_1 and hence must be some v_j where $j \neq k$ (say, with $j > k$). Deletion of v_j from the set S_3 gives the set

$S_4 = \{w_2, w_1, v_1, v_2, \dots, v_{k-1}, v_{k+1}, \dots, v_{j-1}, v_{j+1}, \dots, v_n\}$ of n vectors spanning V .

In this process, at each step we insert one vector from $\{w_1, w_2, \dots, w_m\}$ and delete one vector from $\{v_1, v_2, \dots, v_n\}$.

If $m > n$ after repeating this process n times, we arrive at the set $\{w_n, w_{n-1}, \dots, w_1\}$ which spans V .

Hence w_{n+1} is a linear combination of w_1, w_2, \dots, w_n . Hence $\{w_1, w_2, \dots, w_n, w_{n+1}, \dots, w_m\}$ is linearly dependent which is a contradiction.

Hence $m \leq n$.

Theorem 5.19. Any two bases of a finite dimensional vector space V have the same number of elements.

Proof. Since V is finite dimensional, it has a basis say $S = \{v_1, v_2, \dots, v_n\}$.

Let $S' = \{w_1, w_2, \dots, w_m\}$ be any other basis for V .

Now, $L(S) = V$ and S' is a set of m linearly independent vectors. Hence by Theorem 5.18, $m \leq n$.

Also, since $L(S') = V$ and S is a set of n linearly independent vectors, $n \leq m$. Hence $m = n$.

Definition. Let V be a finite dimensional vector space over a field F . The number of elements in any basis of V is called the *dimension* of V and is denoted by $\dim V$.

Examples

1. $\dim V_n(\mathbf{R}) = n$, since $\{e_1, e_2, \dots, e_n\}$ is a basis of $V_n(\mathbf{R})$.
2. $M_2(\mathbf{R})$ is a vector space of dimension 4 over \mathbf{R} since $\left\{ \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix}, \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} \right\}$ is a basis for $M_2(\mathbf{R})$.
3. \mathbf{C} is a vector space of dimension 2 over \mathbf{R} since $\{1, i\}$ is a basis for \mathbf{C} .
4. Let V be the set of all polynomials of degree $\leq n$ in $\mathbf{R}[x]$. V is a vector space over \mathbf{R} having dimension $n + 1$, since $\{1, x, x^2, \dots, x^n\}$ is a basis for V .

Theorem 5.20. Let V be a vector space of dimension n . Then

- (i) any set of m vectors where $m > n$ is linearly dependent.
- (ii) any set of m vectors where $m < n$ cannot span V .

Proof. (i) Let $S = \{v_1, v_2, \dots, v_n\}$ be a basis for V . Hence $L(S) = V$.

Let S' be any set consisting of m vectors where $m > n$. Suppose S' is linearly independent. Since S spans V by Theorem 5.18, $m \leq n$ which is a contradiction.

Hence S' is linearly dependent.

- (ii) Let S' be a set consisting of m vectors where $m < n$. Suppose $L(S') = V$.

Now, $S = \{v_1, v_2, \dots, v_n\}$ is a basis for V and hence linearly independent. Hence by Theorem 5.18 $n \leq m$ which is a contradiction. Hence S' cannot span V .

Theorem 5.21. Let V be a finite dimensional vector space over a field F . Any linearly independent set of vectors in V is part of a basis.

Proof. Let $S = \{v_1, v_2, \dots, v_r\}$ be a linearly independent set of vectors.

If $L(S) = V$ then S itself is a basis.

If $L(S) \neq V$, choose an element $v_{r+1} \in V - L(S)$.

Now, consider $S_1 = \{v_1, v_2, \dots, v_r, v_{r+1}\}$.

We shall prove that S_1 is linearly independent by showing that no vector in S_1 is a linear combination of the preceding vectors. (refer theorem 5.14).

Since $\{v_1, v_2, \dots, v_r\}$ is linearly independent, v_i where $1 \leq i \leq r$ is not a linear combination of the preceding vectors.

Also $v_{r+1} \notin L(S)$ and hence v_{r+1} is not a linear combination of v_1, v_2, \dots, v_r .

Hence S_1 is linearly independent.

If $L(S_1) = V$, then S_1 is a basis for V . If not we take an element $v_{r+2} \in V - L(S_1)$ and proceed as before. Since the dimension of V is finite, this process

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must stop at a certain stage giving the required basis containing S .

Theorem 5.22. Let V be a finite dimensional vector space over a field F . Let A be a subspace of V . Then there exists a subspace B of V such that $V = A \oplus B$.

Proof. Let $S = \{v_1, v_2, \dots, v_r\}$ be a basis of A .

By theorem 5.21, we can find $w_1, w_2, \dots, w_s \in V$ such that $S' = \{v_1, v_2, \dots, v_r, w_1, w_2, \dots, w_s\}$ is a basis of V .

Now, let $B = L(\{w_1, w_2, \dots, w_s\})$

We claim that $A \cap B = \{0\}$ and $V = A + B$.

Now, let $v \in A \cap B$. Then $v \in A$ and $v \in B$.

$$\begin{aligned} \text{Hence } v &= \alpha_1 v_1 + \dots + \alpha_r v_r \\ &= \beta_1 w_1 + \dots + \beta_s w_s \end{aligned}$$

$$\therefore \alpha_1 v_1 + \dots + \alpha_r v_r - \beta_1 w_1 - \dots - \beta_s w_s = 0.$$

Now, since S' is linearly-independent, $\alpha_i = 0 = \beta_j$ for all i and j .

Hence $v = 0$. Thus $A \cap B = \{0\}$.

Now, let $v \in V$.

$$\begin{aligned} \text{Then } v &= (\alpha_1 v_1 + \dots + \alpha_r v_r) \\ &\quad + (\beta_1 w_1 + \dots + \beta_s w_s) \in A + B. \end{aligned}$$

Hence $A + B = V$ so that $V = A \oplus B$.

Exercises

- Let V be a finite-dimensional vector space. Let A and B be subspaces of V such that $V = A \oplus B$. Then show that $\dim V = \dim A + \dim B$.
- Construct 3 subspaces W_1, W_2, W_3 of a vector space V such that $V = W_1 \oplus W_2 = W_1 \oplus W_3$ but $W_2 \neq W_3$.
- For each of the following subspaces A of $V_3(\mathbf{R})$ find another subspace B such that $A \oplus B = V_3(\mathbf{R})$
 - $A = L\{(1, 1, 0), (0, 1, 1)\}$.
 - $A = L\{(1, 1, 1)\}$.
 - $A = L\{e_1, e_2, e_3\}$.

Definition. Let V be a vector space and

$S = \{v_1, v_2, \dots, v_n\}$ be a set of independent vectors in V . Then S is called a **maximal linearly independent set** if for every $v \in V - S$, the set $\{v, v_1, v_2, \dots, v_n\}$ is linearly dependent.

Definition. Let $S = \{v_1, v_2, \dots, v_n\}$ be a set of vectors in V and let $L(S) = V$. Then S is called a **minimal generating set** if for any $v_i \in S$,

$$L(S - \{v_i\}) \neq V.$$

Theorem 5.23. Let V be a vector space over a field F . Let $S = \{v_1, v_2, \dots, v_n\} \subseteq V$. Then the following are equivalent.

- S is a basis for V .
- S is a maximal linearly independent set.
- S is a minimal generating set.

Proof. (i) \Rightarrow (ii) Let $S = \{v_1, v_2, \dots, v_n\}$ be a basis for V . Then by theorem 5.20 any $n + 1$ vectors in V are linearly dependent and hence S is a maximal linearly independent set.

(ii) \Rightarrow (i) Let $S = \{v_1, v_2, \dots, v_n\}$ be a maximal linearly independent set. Now to prove that S is a basis for V we shall show that $L(S) = V$.

Obviously $L(S) \subseteq V$.

Now, let $v \in V$.

If $v \in S$, then $v \in L(S)$. (since $S \subseteq L(S)$)

If $v \notin S$, $S' = \{v_1, v_2, \dots, v_n, v\}$ is a linearly dependent set (since S is a maximal linearly independent set)

\therefore There exists a vector in S' which is a linear combination of the preceding vectors.

Since v_1, v_2, \dots, v_n are linearly independent, this vector must be v . Thus v is a linear combination of v_1, v_2, \dots, v_n . Therefore $v \in L(S)$.

Hence $V \subseteq L(S)$. Thus $V = L(S)$.

(i) \Rightarrow (iii) Let $S = \{v_1, v_2, \dots, v_n\}$ be a basis. Then $L(S) = V$.

If S is not minimal, there exists $v_i \in S$ such that $L(S - \{v_i\}) = V$.

Since S is linearly independent, $S - \{v_i\}$ is also linearly independent. Thus $S - \{v_i\}$ is a basis consisting of $n - 1$ elements which is a contradiction.

Hence S is a minimal generating set.

(iii) \Rightarrow (i) Let $S = \{v_1, v_2, \dots, v_n\}$ be a minimal generating set. To prove that S is a basis, we have to show that S is linearly independent.

If S is linearly dependent, there exists a vector which is a linear combination of the preceding vectors.

Clearly $L(S - \{v_k\}) = V$ contradicting the minimality of S .

Thus S is linearly independent and since

$$L(S) = V, S \text{ is a basis for } V.$$

Theorem 5.24. Any vector space of dimension n over a field F is isomorphic to $V_n(F)$.

Proof. Let V be a vector space of dimension n . Let $\{v_1, v_2, \dots, v_n\}$ be a basis for V .

Then we know that if $v \in V$, v can be written uniquely as $v = \alpha_1 v_1 + \alpha_2 v_2 + \dots + \alpha_n v_n$, where $\alpha_i \in F$.

Now, consider the map $f : V \rightarrow V_n(F)$ given by

$$f(\alpha_1 v_1 + \dots + \alpha_n v_n) = (\alpha_1, \alpha_2, \dots, \alpha_n).$$

Clearly f is 1-1 and onto.

Let $v, w \in V$.

Then $v = \alpha_1 v_1 + \dots + \alpha_n v_n$ and

$$w = \beta_1 v_1 + \dots + \beta_n v_n.$$

$$\begin{aligned} f(v+w) &= f[(\alpha_1 + \beta_1)v_1 + \dots + (\alpha_n + \beta_n)v_n] \\ &= ((\alpha_1 + \beta_1), (\alpha_2 + \beta_2), \dots, (\alpha_n + \beta_n)) \end{aligned}$$

$$= (\alpha_1, \alpha_2, \dots, \alpha_n) + (\beta_1, \beta_2, \dots, \beta_n)$$

$$= f(v) + f(w).$$

$$\text{Also } f(\alpha v) = f(\alpha \alpha_1 v_1 + \dots + \alpha \alpha_n v_n)$$

$$= (\alpha \alpha_1, \alpha \alpha_2, \dots, \alpha \alpha_n)$$

$$= \alpha(\alpha_1, \alpha_2, \dots, \alpha_n)$$

$$= \alpha f(v).$$

Hence f is an isomorphism of V to $V_n(F)$.

Corollary. Any two vector spaces of the same dimension over a field F are isomorphic, for, if the vector spaces are of dimension n , each is isomorphic to $V_n(F)$ and hence they are isomorphic.

Theorem 5.25. Let V and W be vector spaces over a field F . Let $T : V \rightarrow W$ be an isomorphism. Then T maps a basis of V onto a basis of W .

Proof. Let $\{v_1, v_2, \dots, v_n\}$ be a basis for V .

We shall prove that $T(v_1), T(v_2), \dots, T(v_n)$ are linearly independent and that they span W .

$$\text{Now, } \alpha_1 T(v_1) + \alpha_2 T(v_2) + \dots + \alpha_n T(v_n) = 0.$$

$$\Rightarrow T(\alpha_1 v_1) + T(\alpha_2 v_2) + \dots + T(\alpha_n v_n) = 0.$$

$$\Rightarrow T(\alpha_1 v_1 + \alpha_2 v_2 + \dots + \alpha_n v_n) = 0.$$

$$\Rightarrow \alpha_1 v_1 + \alpha_2 v_2 + \dots + \alpha_n v_n = 0 \quad (\text{since } T \text{ is 1-1})$$

$$\Rightarrow \alpha_1 = \alpha_2 = \dots = \alpha_n = 0$$

(since v_1, v_2, \dots, v_n are linearly independent).

$\therefore T(v_1), T(v_2), \dots, T(v_n)$ are linearly independent.

Now, let $w \in W$. Then since T is onto, there exists a vector $v \in V$ such that $T(v) = w$.

$$\text{Let } v = \alpha_1 v_1 + \dots + \alpha_n v_n.$$

$$\text{Then } w = T(v)$$

$$= T(\alpha_1 v_1 + \dots + \alpha_n v_n).$$

$$= \alpha_1 T(v_1) + \dots + \alpha_n T(v_n).$$

Thus w is a linear combination of the vectors

$$T(v_1), \dots, T(v_n).$$

$\therefore T(v_1), \dots, T(v_n)$ span W and hence is a basis for W .

Corollary. Two finite dimensional vector spaces V and W over a field F are isomorphic iff they have the same dimension.

Theorem 5.26. Let V and W be finite dimensional vector spaces over a field F . Let $\{v_1, v_2, \dots, v_n\}$ be a basis for V and let w_1, w_2, \dots, w_n be any n vectors in W (not necessarily distinct). Then there exists a unique linear transformation $T : V \rightarrow W$ such that $T(v_i) = w_i, i = 1, 2, \dots, n$.

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Proof. Let $v = \alpha_1 v_1 + \alpha_2 v_2 + \dots + \alpha_n v_n \in V$.

We define $T(v) = \alpha_1 w_1 + \alpha_2 w_2 + \dots + \alpha_n w_n$.

Now, let $x, y \in V$.

Let $x = \alpha_1 v_1 + \dots + \alpha_n v_n$ and

$y = \beta_1 v_1 + \dots + \beta_n v_n$.

$$\therefore x + y = (\alpha_1 + \beta_1)v_1 + \dots + (\alpha_n + \beta_n)v_n.$$

$$\therefore T(x + y) = (\alpha_1 + \beta_1)w_1 + \dots + (\alpha_n + \beta_n)w_n$$

$$= (\alpha_1 w_1 + \dots + \alpha_n w_n) +$$

$$(\beta_1 w_1 + \dots + \beta_n w_n)$$

$$= T(x) + T(y).$$

Similarly $T(\alpha x) = \alpha T(x)$.

Hence T is a linear transformation.

Also $v_1 = 1v_1 + 0v_2 + \dots + 0v_n$.

Hence $T(v_1) = 1w_1 + 0w_2 + 0w_n = w_1$.

Similarly $T(v_i) = w_i$ for all $i = 1, 2, \dots, n$.

Now, to prove the uniqueness, let $T' : V \rightarrow W$ be any other linear transformation such that $T'(v_i) = w_i$.

Let $v = \alpha_1 v_1 + \dots + \alpha_n v_n \in V$

$$T'(v) = \alpha_1 T'(v_1) + \dots + \alpha_n T'(v_n)$$

$$= \alpha_1 w_1 + \dots + \alpha_n w_n = T(v).$$

Hence $T = T'$.

Remark. The above theorem shows that a linear transformation is completely determined by its values on the elements of a basis.

Theorem 5.27. Let V be a finite dimensional vector space over a field F . Let W be a subspace of V . Then

(i) $\dim W \leq \dim V$.

(ii) $\dim \frac{V}{W} = \dim V - \dim W$.

Proof.

(i) Let $S = \{w_1, w_2, \dots, w_m\}$ be a basis for W . Since W is a subspace of V , S is a part of a basis for V .

Hence $\dim W \leq \dim V$.

(ii) Let $\dim V = n$ and $\dim W = m$.

Let $S = \{w_1, w_2, \dots, w_m\}$ be a basis for W . Clearly S is a linearly independent set of vectors in V .

Hence S is a part of a basis in V . Let $\{w_1, w_2, \dots, w_m, v_1, v_2, \dots, v_r\}$ be a basis for V . Then $m + r = n$.

Now, we claim $S' = \{W + v_1, W + v_2, \dots, W + v_r\}$ is a basis for $\frac{V}{W}$.

$$\alpha_1(W + v_1) + \alpha_2(W + v_2) + \dots + \alpha_r(W + v_r) = W +$$

$$\Rightarrow (W + \alpha_1 v_1) + (W + \alpha_2 v_2) + \dots + (W + \alpha_r v_r) = W$$

$$\Rightarrow W + \alpha_1 v_1 + \alpha_2 v_2 + \dots + \alpha_r v_r = W$$

$$\Rightarrow \alpha_1 v_1 + \alpha_2 v_2 + \dots + \alpha_r v_r \in W.$$

Now, since $\{w_1, w_2, \dots, w_m\}$ is a basis for W

$$\alpha_1 v_1 + \dots + \alpha_r v_r = \beta_1 w_1 + \dots + \beta_m w_m.$$

$$\therefore \alpha_1 v_1 + \dots + \alpha_r v_r - \beta_1 w_1 - \dots - \beta_m w_m = 0.$$

$$\therefore \alpha_1 = \alpha_2 = \dots = \alpha_r = \beta_1 = \beta_2 = \dots = \beta_m = 0$$

$\therefore S'$ is a linearly independent set.

Now, let $W + v \in \frac{V}{W}$.

Let $v = \alpha_1 v_1 + \dots + \alpha_r v_r + \beta_1 w_1 + \dots + \beta_m w_m$.

Then $W + v = W + (\alpha_1 v_1 + \dots + \alpha_r v_r$

$$+ \beta_1 w_1 + \dots + \beta_m w_m)$$

$$= W + (\alpha_1 v_1 + \dots + \alpha_r v_r)$$

$$(since \beta_1 w_1 + \dots + \beta_m w_m \in W)$$

$$= (W + \alpha_1 v_1) + \dots + (W + \alpha_r v_r)$$

$$= \alpha_1(W + v_1) + \dots + \alpha_r(W + v_r).$$

Hence S' spans $\frac{V}{W}$ so that S' is a basis for $\frac{V}{W}$.

$$\therefore \dim \frac{V}{W} = r = n - m$$

$$= \dim V - \dim W.$$

Theorem 5.28. Let V be a finite-dimensional vector space over a field F . Let A and B be subspaces of V .

Then $\dim(A + B) = \dim A + \dim B - \dim(A \cap B)$

Proof. A and B are subspaces of V . Hence $A \cap B$ is subspace of V .

Let $\dim(A \cap B) = r$.

Let $S = \{v_1, v_2, \dots, v_r\}$ be a basis for $A \cap B$

Since $A \cap B$ is a subspace of A and B , S is a part of a basis for A and B .

Let $\{v_1, v_2, \dots, v_r, u_1, u_2, \dots, u_s\}$ be a basis for A and $\{v_1, v_2, \dots, v_r, w_1, w_2, \dots, w_t\}$ be a basis for B .

We shall prove that $S' = \{v_1, \dots, v_r, u_1, \dots, u_s, w_1, \dots, w_t\}$ is a basis for $A + B$.

Let $\alpha_1 v_1 + \dots + \alpha_r v_r + \beta_1 u_1 + \dots + \beta_s u_s + \gamma_1 w_1 + \dots + \gamma_t w_t = 0$.

Then $\beta_1 u_1 + \dots + \beta_s u_s = -(\gamma_1 w_1 + \dots + \gamma_t w_t) - (\alpha_1 v_1 + \dots + \alpha_r v_r) \in B$

Hence $\beta_1 u_1 + \dots + \beta_s u_s \in B$.

Also $\beta_1 u_1 + \dots + \beta_s u_s \in A$.

Hence $\beta_1 u_1 + \dots + \beta_s u_s \in A \cap B$.

$$\therefore \beta_1 u_1 + \dots + \beta_s u_s = \delta_1 v_1 + \dots + \delta_r v_r.$$

$$\therefore \beta_1 u_1 + \dots + \beta_s u_s - \delta_1 v_1 - \dots - \delta_r v_r = 0.$$

$$\therefore \beta_1 = \dots = \beta_s = \delta_1 = \dots = \delta_r = 0$$

(since $\{u_1, \dots, u_s, v_1, \dots, v_r\}$ is linearly independent)

Similarly we can prove $\gamma_1 = \gamma_2 = \dots = \gamma_t = 0$.

$$\therefore \alpha_i = \beta_j = \gamma_k = 0 \text{ for } 1 \leq i \leq r,$$

$$1 \leq j \leq s; 1 \leq k \leq t$$

Thus S' is a linearly independent set.

Clearly S' spans $A + B$.

$\therefore S'$ is a basis for $A + B$.

Hence $\dim(A + B) = r + s + t$.

Also $\dim A = r + s$; $\dim B = r + t$ and

$\dim(A \cap B) = r$.

$$\begin{aligned} \therefore \dim A + \dim B - \dim A \cap B &= (r+s) + (r+t) - r \\ &= r + s + t \\ &= \dim(A + B). \end{aligned}$$

Aliter. By theorem 5.7, $\frac{A+B}{A} = \frac{B}{A \cap B}$.

$$\text{Hence } \dim \left[\frac{A+B}{A} \right] = \dim \left[\frac{B}{A \cap B} \right].$$

$$\therefore \dim(A + B) - \dim A = \dim B - \dim(A \cap B).$$

$$\therefore \dim(A + B) = \dim A + \dim B - \dim(A \cap B).$$

Corollary. If $V = A \oplus B$, $\dim V = \dim A + \dim B$.

Proof. $V = A \oplus B \Rightarrow A + B = V$ and $A \cap B = \{0\}$

$$\therefore \dim(A \cap B) = 0.$$

Hence $\dim V = \dim A + \dim B$.

Exercises

1. Find the dimension of the subspace spanned by the following vectors in $V_3(\mathbf{R})$.

(a) $(1, 1, 1), (-1, -1, -1)$.

(b) $(1, 0, 2), (2, 0, 1), (1, 0, 1)$

(c) $(1, 2, -3), (0, 0, 1), (-1, 2, 1)$.

(d) $(1, 1, 2), (-1, 1, 0)$.

2. Find the dimension of the subspace spanned by the following vectors in $V_4(\mathbf{R})$

(a) e_1, e_2, e_3, e_4

(b) e_1, e_2

(c) e_1, e_2, e_3

(d) e_1

3. In $V_3(\mathbf{R})$, find $\dim(A + B)$ and $\dim(A \cap B)$ where

(a) A is the subspace spanned by $(1, 1, 1)$ and B is the subspace spanned by $(-1, -1, -1)$

(b) A is the subspace spanned by $(1, 1, 1)$ and B is the subspace spanned by $(1, 2, 1)$.

- (c) A is the subspace spanned by $(1, 1, 1)$ and $(1, 2, 1)$ and B is the subspace spanned by $(0, 0, 1)$.
- (d) A is the subspace spanned by $(1, 1, 1)$ and $(1, 2, 1)$ and B is the subspace spanned by $(1, -1, 1)$ and $(-1, 1, -1)$.
4. Let V_1 and V_2 be subspaces of V such that $V_1 \cap V_2$ is the zero space. Prove that $\dim V_1 + \dim V_2 \leq \dim V$.
5. Let V_1 and V_2 be subspaces of V such that every vector $v \in V$ can be represented as $v = v_1 + v_2$ where $v_1 \in V_1$ and $v_2 \in V_2$. Prove that $\dim V_1 + \dim V_2 \geq \dim V$.
6. If A and B are finite dimensional subspaces of V such that $A \subseteq B$ and $\dim A = \dim B$ then show that $A = B$.
7. Let S be a subspace of a finite-dimensional vector space V . If $\dim V = \dim S$ then prove that $S = V$.
8. Let W_1 and W_2 be two subspaces of a finite-dimensional vector space V . If $\dim V = \dim W_1 + \dim W_2$ and $W_1 \cap W_2 = \{0\}$ prove that, $V = W_1 \oplus W_2$.

Answers.

1. (a) 1 (b) 2 (c) 3 (d) 2
 2. (a) 4 (b) 2 (c) 3 (d) 1
 3. (a) 1; 1 (b) 2; 0 (c) 3; 0 (d) 3; 0

5.7. Rank and Nullity

Definition. Let $T : V \rightarrow W$ be a linear transformation. Then the dimension of $T(V)$ is called the *rank* of T . The dimension of $\ker T$ is called the *nullity* of T .

Theorem 5.29. Let $T : V \rightarrow W$ be a linear transformation. Then $\dim V = \text{rank } T + \text{nullity } T$.

Proof. We know that $V/\ker T = T(V)$.

$$\therefore \dim V - \dim(\ker T) = \dim(T(V))$$

$$\therefore \dim V - \text{nullity } T = \text{rank } T$$

$$\therefore \dim V = \text{nullity } T + \text{rank } T$$

Note. $\ker T$ is also called *null space* of T .

Example. Let V denote the set of all polynomials of degree $\leq n$ in $\mathbb{R}[x]$. Let $T : V \rightarrow V$ be defined by $T(f) = \frac{df}{dx}$. We know that T is a linear transformation. Since $\frac{df}{dx} = 0 \Leftrightarrow f$ is constant, $\ker T$ consists of all constant polynomials. The dimension of this subspace of V is 1. Hence *nullity* T is 1. Since $\dim V = n + 1$, $\text{rank } T = n$.

Exercises

- Find the *rank* and *nullity* of the linear transformations given in section 5.3.
- Let V be a finite-dimensional vector space over a field F . Let $T : V \rightarrow V$ be a linear transformation such that $\text{rank } T = \text{nullity } T$. Show that $\dim V$ is even. Give an example of such a transformation.

Answers.

- $\text{nullity } T = \dim V$; $\text{rank } T = 0$.
 - $\text{nullity } T = 0$; $\text{rank } T = \dim V$.
 - $\text{nullity } T = \dim W$;
 $\text{rank } T = \dim V - \dim W$.
 - $\text{nullity } T = 2$; $\text{rank } T = 1$;
 - $\text{nullity } T = 1$; $\text{rank } T = n$.
 - $\text{nullity } T = 0$; $\text{rank } T = n + 1$.

Definition. A linear transformation $T : V \rightarrow W$ is called *non-singular* if T is 1-1; otherwise T is called *singular*.

Exercises

- Let V and W be finite dimensional vector spaces over a field F and $\dim V > \dim W$. Then show that any linear transformation $T : V \rightarrow W$ is singular.
- Let V be a finite-dimensional vector space over a field F . Then any non-singular linear transformation $T : V \rightarrow V$ is onto.
- Let $T : V \rightarrow W$ be a linear transformation. Show that T is a non-singular iff $\text{rank } T = \dim V$.

4. Let $T_1 : V \rightarrow V$ and $T_2 : V \rightarrow V$ be linear transformations. Prove that

- (a) $rank(T_2T_1) \leq rank T_2$.
- (b) $nullity(T_2T_1) \geq nullity T_1$.
- (c) $rank(T_2T_1) = rank T_2$ iff T_1 is non-singular.

5. Let $T : V \rightarrow W$ be a linear transformation which is both 1-1 and onto. Show that $T^{-1} : W \rightarrow V$ is a linear transformation.

6. Determine which of the following statements are true and which are false.

- (a) If $T : V \rightarrow W$ is a linear transformation then
 - (i) $rank T \leq dim V$
 - (ii) $nullity T \leq dim V$.
 - (iii) $rank T \leq dim W$.
 - (iv) If T is onto $rank T = dim W$.
 - (v) If T is non-singular $rank T = dim V$
 - (vi) $rank T = dim V \Rightarrow nullity T = 0$.

- (b) Every linear transformation $T : V_4(\mathbb{R}) \rightarrow V_3(\mathbb{R})$ is singular.
- (c) If $T : V \rightarrow W$ is non-singular and $\{v_1, \dots, v_n\}$ is a basis then $\{T(v_1), \dots, T(v_n)\}$ is a basis for W .

Answers.

- 6.(a) (i) T (ii) T (iii) T (iv) T (v) T (vi) T
- (b) T (c) F.

5.8. Matrix of a Linear Transformation

Let V and W be finite dimensional vector spaces over a field F . Let $dim V = m$ and $dim W = n$. Fix an ordered basis $\{v_1, v_2, \dots, v_m\}$ for V and an ordered basis $\{w_1, w_2, \dots, w_n\}$ for W .

Let $T : V \rightarrow W$ be a linear transformation. We have seen that T is completely specified by the elements $T(v_1), T(v_2), \dots, T(v_m)$. Now, let

$$\begin{aligned} T(v_1) &= a_{11}w_1 + a_{12}w_2 + \dots + a_{1n}w_n \\ T(v_2) &= a_{21}w_1 + a_{22}w_2 + \dots + a_{2n}w_n \quad (1) \\ &\dots \\ T(v_m) &= a_{m1}w_1 + a_{m2}w_2 + \dots + a_{mn}w_n \end{aligned}$$

Hence $T(v_1), T(v_2), \dots, T(v_m)$ are completely specified by the mn elements a_{ij} of the field F . These a_{ij} can be conveniently arranged in the form of m rows and n columns as follows.

$$\begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \dots & \dots & \dots & \dots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{pmatrix}$$

Such an array of mn elements of F arranged in m rows and n columns is known as $m \times n$ matrix over the field F and is denoted by (a_{ij}) . Thus to every linear transformation T there is associated with it an $m \times n$ matrix over F . Conversely any $m \times n$ matrix over F defines a linear transformation $T : V \rightarrow W$ given by the formula (1).

Note. The $m \times n$ matrix which we have associated with a linear transformation $T : V \rightarrow W$ depends on the choice of the basis for V and W .

For example, consider the linear transformation $T : V_2(\mathbb{R}) \rightarrow V_2(\mathbb{R})$ given by $T(a, b) = (a, a + b)$. Choose $\{e_1, e_2\}$ as a basis both for the domain and the range.

$$\begin{aligned} \text{Then } T(e_1) &= (1, 1) = e_1 + e_2 \\ T(e_2) &= (0, 1) = e_2. \end{aligned}$$

Hence the matrix representing T is $\begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix}$

Now, we choose $\{e_1, e_2\}$ as a basis for the domain and $\{(1, 1), (1, -1)\}$ as a basis for the range.

Let $w_1 = (1, 1)$ and $w_2 = (1, -1)$.

Then $T(e_1) = (1, 1) = w_1$,

and $T(e_2) = (0, 1) = (1/2)w_1 - (1/2)w_2$.

Hence the matrix representing T is $\begin{pmatrix} 1 & 0 \\ 1/2 & -1/2 \end{pmatrix}$

Solved problems

Problem 1. Obtain the matrix representing the linear transformation $T : V_3(\mathbf{R}) \rightarrow V_3(\mathbf{R})$ given by $T(a, b, c) = (3a, a - b, 2a + b + c)$ w.r.t. the standard basis $\{e_1, e_2, e_3\}$.

Solution.

$$T(e_1) = T(1, 0, 0) = (3, 1, 2) = 3e_1 + e_2 + 2e_3$$

$$T(e_2) = T(0, 1, 0) = (0, -1, 1) = -e_2 + e_3$$

$$T(e_3) = T(0, 0, 1) = (0, 0, 1) = e_3$$

Thus the matrix representing T is $\begin{pmatrix} 3 & 1 & 2 \\ 0 & -1 & 1 \\ 0 & 0 & 1 \end{pmatrix}$

Problem 2. Find the linear transformation

$T : V_3(\mathbf{R}) \rightarrow V_3(\mathbf{R})$ determined by the matrix $\begin{pmatrix} 1 & 2 & 1 \\ 0 & 1 & 1 \\ -1 & 3 & 4 \end{pmatrix}$ w.r.t. the standard basis $\{e_1, e_2, e_3\}$.

Solution.

$$T(e_1) = e_1 + 2e_2 + e_3 = (1, 2, 1).$$

$$T(e_2) = 0e_1 + e_2 + e_3 = (0, 1, 1)$$

$$T(e_3) = -e_1 + 3e_2 + 4e_3 = (-1, 3, 4).$$

Now, $(a, b, c) = a(1, 0, 0) + b(0, 1, 0) + c(0, 0, 1)$

$$= ae_1 + be_2 + ce_3.$$

$$\therefore T(a, b, c) = T(ae_1 + be_2 + ce_3)$$

$$= aT(e_1) + bT(e_2) + cT(e_3)$$

$$= a(1, 2, 1) + b(0, 1, 1) + c(-1, 3, 4).$$

$$\therefore T(a, b, c) = (a - c, 2a + b + 3c, a + b + 4c)$$

This is the required linear transformation.

Exercises

1. Obtain the matrices for the following linear transformations:

(a) $T : V_2(\mathbf{R}) \rightarrow V_2(\mathbf{R})$ given by $T(a, b) = (-b, a)$ w.r.t.

(i) standard basis

(ii) the basis $\{(1, 2), (1, -1)\}$ for both domain and range.

(b) $T : V_3(\mathbf{R}) \rightarrow V_2(\mathbf{R})$ given by $T(a, b, c) = (a + b, 2c - a)$ w.r.t.

(i) standard basis

(ii) $\{(1, 0, -1), (1, 1, 1), (1, 0, 0)\}$ as a basis for $V_3(\mathbf{R})$ and $\{(0, 1), (1, 0)\}$ for $V_2(\mathbf{R})$.

(c) $T : V_3(\mathbf{R}) \rightarrow V_3(\mathbf{R})$ given by

$$T(a, b, c) = (3a + c, -2a + b, a + 2b + 4c) \text{ w.r.t.}$$

(i) the standard basis

(ii) the basis $\{(1, 0, 1), (-1, 2, 1), (2, 1, 1)\}$ for both domain and range.

(d) Let V be the set of all polynomials of degree $\leq n$ in $\mathbf{R}[x]$.

$$T : V \rightarrow V \text{ defined by } T(f) = \frac{df}{dx} \text{ w.r.t. the basis } \{1, x, x^2, \dots, x^n\}.$$

2. Obtain the linear transformation determined by the following matrices

(a) $T : V_2(\mathbf{R}) \rightarrow V_2(\mathbf{R})$ given by $\begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix}$ w.r.t. the standard basis.

(b) $T : V_3(\mathbf{R}) \rightarrow V_3(\mathbf{R})$ given by $\begin{pmatrix} a & b & c \\ b & c & a \\ c & a & b \end{pmatrix}$ w.r.t. the standard basis.

(c) $T : V_2(\mathbf{R}) \rightarrow V_3(\mathbf{R})$ given by $\begin{pmatrix} 2 & 1 & -1 \\ 1 & 1 & -1 \end{pmatrix}$ w.r.t. the standard basis.

Answers.

1. (a) (i) $\begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$ (ii) $\begin{pmatrix} -1/3 & -5/3 \\ 2/3 & 1/3 \end{pmatrix}$

(b) (i) $\begin{pmatrix} 1 & 1 \\ 1 & 0 \\ 0 & 2 \end{pmatrix}$ (ii) $\begin{pmatrix} -3 & 1 \\ 1 & 2 \\ -1 & 1 \end{pmatrix}$

(c) (i) $\begin{pmatrix} 3 & 2 & -1 \\ 0 & 1 & 2 \\ 1 & 0 & 4 \end{pmatrix}$

(ii) $\begin{pmatrix} 17/4 & -3/4 & -1/2 \\ 35/4 & 15/4 & -7/2 \\ 17/2 & -3/2 & 0 \end{pmatrix}$

(d) $\begin{pmatrix} 0 & 0 & 0 & \dots & 0 & 0 \\ 1 & 0 & 0 & \dots & 0 & 0 \\ 0 & 2 & 0 & \dots & 0 & 0 \\ 0 & 0 & 3 & \dots & 0 & 0 \\ \dots & \dots & \dots & \dots & \dots & \dots \\ 0 & 0 & 0 & \dots & n & 0 \end{pmatrix}$

2. (a) $T(a, b) = (a \cos \theta + b \sin \theta, -a \sin \theta + b \cos \theta)$

(b) $T(x, y, z) = (ax + by + cz, bx + cy + az, cx + ay + bz)$

(c) $T(a, b) = (2a + b, a + b, -a - b)$.

Definition. Let $A = (a_{ij})$ and $B = (b_{ij})$ be two $m \times n$ matrices. We define the *sum* of these two matrices by $A + B = (a_{ij} + b_{ij})$.

Note that we have defined addition only for two matrices having the same number of rows and the same number of columns.

Definition. Let $A = (a_{ij})$ be an arbitrary matrix over a field F . Let $\alpha \in F$. We define $\alpha A = (\alpha a_{ij})$.

Theorem 5.30. The set $M_{m \times n}(F)$ of all $m \times n$ matrices over the field F is a vector space of dimension mn over F under matrix addition and scalar multiplication defined above.

Proof. Let $A = (a_{ij})$ and $B = (b_{ij})$ be two $m \times n$ matrices over the field F . The addition of $m \times n$ matrices is a binary operation which is both commutative and associative. The $m \times n$ matrix whose entries are 0 is the *identity matrix* and $(-a_{ij})$ is the *inverse matrix* of (a_{ij}) . Thus the set of all $m \times n$ matrices over the field F is an *abelian group* with respect to addition. The verification of the following axioms are straight forward.

(a) $\alpha(A + B) = \alpha A + \alpha B$

(b) $(\alpha + \beta)A = \alpha A + \beta A$

(c) $(\alpha\beta)A = \alpha(\beta A)$

(d) $1A = A$.

Hence the set of all $m \times n$ over F is a vector space over F .

Now, we shall prove that the dimension of this vector space is mn . Let E_{ij} be the matrix with entry 1 in the (i, j) th place and 0 in the other places. We have mn matrices of this form. Also any matrix $A = (a_{ij})$ can be written as $A = \sum a_{ij} E_{ij}$. Hence A is a linear combination of the matrices E_{ij} . Further these mn matrices E_{ij} are linearly independent. Hence these mn matrices form a basis for the space of all $m \times n$ matrices. Therefore the dimension of the vector space is mn .

Theorem 5.31. Let V and W be two finite dimensional vector spaces over a field F . Let $\dim V = m$ and $\dim W = n$. Then $L(V, W)$ is a vector space of dimension mn over F .

Proof. By theorem 5.8, $L(V, W)$ is a vector space over F . Now, we shall prove that the vector space $L(V, W)$ is isomorphic to the vector space $M_{m \times n}(F)$. Since $M_{m \times n}(F)$ is of dimension mn , it follows that $L(V, W)$ is also of dimension mn .

Fix a basis $\{v_1, v_2, \dots, v_m\}$ for V and a basis $\{w_1, w_2, \dots, w_n\}$ for W .

We know that any linear transformation

$T \in L(V, W)$ can be represented by an $m \times n$ matrix over F .