

Chapter 3

Linear Mappings

In this chapter a class of relationships between vector spaces called **linear mappings** are examined.

3.1 General Mappings

A mapping between two sets A, B denoted $M : A \rightarrow B$ is a rule assigning to each element $a \in A$ an element $b \in B$ called the **image** of a under the mapping M written $b = M(a)$. The set A is the **domain** of M and the set B its **codomain**.

We define the **range** of M (or the image of A) as the set of images of all elements of A and is denoted as $\mathcal{R}(M) = \{b \in B / b = M(a)\}$. Clearly $\mathcal{R}(M)$ is a subset of B .

The rule of assigning a student number to every student is a simple example of a mapping. The real function $f : \mathbb{R} \rightarrow \mathbb{R}$ defined by $f(x) = x^2$ is another example of a mapping.

Mappings may enjoy some useful properties defined below:

Definition 35 A mapping $M : A \rightarrow B$ is **one-to-one** or **injective** if and only if **distinct elements** have **distinct images**.

This can be expressed as follows: M one-to-one \Leftrightarrow For all $a \neq a'$ we have $M(a) \neq M(a')$ \Leftrightarrow For all $M(a) = M(a')$ we have $a = a'$.

A one-to-one mapping may also be termed as an invertible mapping: Every element of A has a **unique** image allowing to identifying it.

Example 36 Consider $f : \mathbb{R} \rightarrow \mathbb{R}$ defined by $f(x) = ax + b$ where a, b are real constants. Let $x_1 \neq x_2$ and consider their images $f(x_1) = ax_1 + b$ and $f(x_2) = ax_2 + b$ leading to $f(x_1) - f(x_2) = a(x_1 - x_2)$. Clearly for $x_1 \neq x_2$ or $x_1 - x_2 \neq 0$, we have $f(x_1) \neq f(x_2)$ indicating that f is one-to-one provided $a \neq 0$.

Example 37 Consider now $f : \mathbb{R} \rightarrow \mathbb{R}$ defined by $f(x) = x^2$ which is not since $f(1) = f(-1)$.

Definition 38 A mapping $M : A \rightarrow B$ is **unto** or **surjective** if and only if every element of B is an image of an element of A .

This can be expressed as: M unto \Leftrightarrow For any $b \in B$, there exists $a \in A$ such that $b = M(a)$. It essentially states that **all** elements of B are **images** leading to $B = \mathcal{R}(M)$

Example 39 Consider again $f : \mathbb{R} \rightarrow \mathbb{R}$ defined by $f(x) = ax + b$. For any $y \in \mathbb{R}$, we can find $x \in \mathbb{R}$ such that $y = ax + b$, hence $x = \frac{y-b}{a}$ for $a \neq 0$ indicating that the mapping is onto. On the other hand the mapping $f(x) = x^2$ is not since strictly negative real numbers cannot be images.

Definition 40 A mapping $M : A \rightarrow B$ is **bijective** if and only if it is both injective and surjective.

For a bijective mapping every element of the domain has a unique image and every element of the codomain is the image of a unique element from the domain.

In linear algebra, we are interested in a special class of mappings: the so-called **linear mappings**.

3.2 Linear Mappings

Consider the two vector spaces $V_n(F)$ and $W_m(F)$ and a mapping $L : V_n(F) \rightarrow W_m(F)$

Definition 41 The mapping $L : V_n(F) \rightarrow W_m(F)$ is said to be **linear** if and only if we have:

- i) For all $v_1, v_2 \in V_n(F)$, we have $L(v_1 + v_2) = L(v_1) + L(v_2)$
- ii) For all $v \in V_n(F)$ and all $\alpha \in F$, we have $L(\alpha v) = \alpha L(v)$

The first property is called **additivity** while the second is called **homogeneity**. The additivity property requires that the image of a sum of vectors be the sum of their images. The homogeneity property requires that the image of a scaled vector be the its scaled image. These properties constitute what is called the **principal of superposition in engineering**.

Both properties can be combined into a single property: L linear if and only if $L(\alpha_1 v_1 + \alpha_2 v_2) = \alpha_1 L(v_1) + \alpha_2 L(v_2)$ requiring that the image of a linear combination of images be the linear combination of their images. This can be written shortly as: L linear $\Leftrightarrow \sum_{i=1}^k L(\alpha_i v_i) = \sum_{i=1}^k \alpha_i L(v_i)$.

Remark 42 A linear mapping where the domain and codomain are identical is called a **linear transformation** as in $T : V_n(F) \rightarrow V_n(F)$

Example 43 Consider the mapping $L : \mathbb{R}^2(\mathbb{R}) \rightarrow \mathbb{R}^2(\mathbb{R})$ defined by $L(x, y) = (0, y)$. To check linearity, consider any $v_1, v_2 \in \mathbb{R}^2(\mathbb{R})$ and any $\alpha_1, \alpha_2 \in \mathbb{R}$, then

$$L(\alpha_1 v_1 + \alpha_2 v_2) = L[\alpha_1(x_1, y_1) + \alpha_2(x_2, y_2)] = L(\alpha_1 x_1 + \alpha_2 x_2, \alpha_1 y_1 + \alpha_2 y_2) = (0, \alpha_1 y_1 + \alpha_2 y_2) = \alpha_1(0, y_1) + \alpha_2(0, y_2) = \alpha_1 L(v_1) + \alpha_2 L(v_2) \text{ establishing that } L \text{ is linear.}$$

This mapping represents an orthogonal projection onto the y -axis of the $y - x$ plane.

Example 44 Consider the vector space $V(\mathbb{R}) = \{V(\mathbb{R}) : \mathbb{R} \rightarrow \mathbb{R} / f \text{ differentiable}\}$ and the mapping $D : V(\mathbb{R}) \rightarrow V(\mathbb{R})$ defined by $D[f] = \frac{df(x)}{dx}$. Let $f_1, f_2 \in V(\mathbb{R})$ and $\alpha_1, \alpha_2 \in \mathbb{R}$ and consider $D[\alpha_1 f_1 + \alpha_2 f_2] = \frac{d}{dx}[\alpha_1 f_1(x) + \alpha_2 f_2(x)]$. It follows that $\frac{d}{dx}[\alpha_1 f_1(x) + \alpha_2 f_2(x)] = \alpha_1 \frac{df_1(x)}{dx} + \alpha_2 \frac{df_2(x)}{dx} = \alpha_1 D[f_1] + \alpha_2 D[f_2]$, hence $D[\alpha_1 f_1 + \alpha_2 f_2] = \alpha_1 D[f_1] + \alpha_2 D[f_2]$ and the mapping is linear.

Example 45 Consider the mapping $L : \mathbb{R}^2(\mathbb{R}) \rightarrow \mathbb{R}^2(\mathbb{R})$ defined by $L(x, y) = (0, xy)$. Clearly $L(\alpha v) = L[\alpha(x, y)] = L(\alpha x, \alpha y) = (0, \alpha^2 xy) \neq \alpha L(v) = \alpha L(x, y) = \alpha(0, xy) = (0, \alpha xy)$ and the mapping is not linear.

A linear mapping is also called a **space homomorphism** and if bijective also, it is called a **space isomorphism**.

Remark 46 From the homogeneity property $L(\alpha v) = \alpha L(v)$ for any $\alpha \in F$ and any $v \in V_n(F)$, then for the particular case $\alpha = 0$ we have $L(0v) = 0L(v)$ or $L(\theta) = \theta$ meaning that a linear mapping always maps the zero vector of $V_n(F)$ into the zero vector of $W_m(F)$.

Consider the linear mapping $L : V_n(F) \rightarrow W_m(F)$ and suppose we select a basis $\{v_1, v_2, \dots, v_n\}$ for $V_n(F)$. If v is any vector of $V_n(F)$, we can express it uniquely as $v = \sum_{i=1}^n \alpha_i v_i$ and its image as $L(v) = \sum_{i=1}^n L(\alpha_i v_i)$ and by linearity $L(v) = \sum_{i=1}^n \alpha_i L(v_i)$ meaning that the image of v is expressed uniquely as a linear combination of the images of the vectors of the selected basis. It follows that the effect of the mapping L is described uniquely by the effect of the mapping on the vectors of any selected basis.

3.3 Fundamental Subspaces

Consider the linear mapping $L : V_n(F) \rightarrow W_m(F)$ and suppose that S is a subspace of $V_n(F)$. Is the image of S i.e. $L(S)$ also a subspace? Clearly $L(S)$ is a nonzero subset of $W_m(F)$, thus we need only check $L(S)$ for closedness of under linear combinations. Let $w_1, w_2 \in L(S)$, hence there exists $v_1, v_2 \in V_n(F)$ such that $w_1 = L(v_1)$ and $w_2 = L(v_2)$. Now $\alpha_1 w_1 + \alpha_2 w_2 = \alpha_1 L(v_1) + \alpha_2 L(v_2)$ and by linearity $\alpha_1 w_1 + \alpha_2 w_2 = L(\alpha_1 v_1 + \alpha_2 v_2)$ implying that $\alpha_1 w_1 + \alpha_2 w_2$ is an image or $\alpha_1 w_1 + \alpha_2 w_2 \in L(S)$. Thus, $L(S)$ is closed under linear combination therefore a subspace. We can thus conclude that under a linear mapping every

subspace of the domain is mapped into a subspace of the codomain or **the image of a subspace is also a subspace**.

Every linear mapping defines two special subspaces called **fundamental subspaces**: one in the domain and one in the codomain.

Definition 47 The range space of the linear mapping $L : V_n(F) \rightarrow W_m(F)$ is defined as $\mathcal{R}(L) = \{w \in W_m(F) / w = L(v) \text{ where } v \in V_n(F)\}$; The dimension of $\mathcal{R}(L)$ is called the rank of L and denoted $\text{rank}(L)$ or $r(L)$.

The range space $\mathcal{R}(L)$ is the image of $V_n(F)$ and since the image of subspace is also a subspace, then $\mathcal{R}(L)$ is a subspace of $W_m(F)$.

Example 48 Consider the linear mapping $D : P_3 \rightarrow P_3$, where $P_3 = \{p(x)/p(x) = a_0 + a_1x + a_2x^2 + a_3x^3\}$ is the space of real polynomials of degree ≤ 3 and defined by $D[p(x)] = \frac{dp(x)}{dx}$. The range space is $\mathcal{R}(L) = \{p'(x)/p'(x) = \frac{dp(x)}{dx}\} = \{p'(x) = a_1 + 2a_2x + 3a_3x^2\}$, hence the space of all real polynomials of degree ≤ 2 . Clearly a basis for this subspace is $\{1, x, x^2\}$ and $r(D) = 3$.

The rank of a mapping is important concept in linear algebra and we are often required to compute it. Since this rank is the dimension of $\mathcal{R}(L)$, it is determined by the number of vectors in any basis of $\mathcal{R}(L)$. In the following we attempt to find bases for it.

Theorem 49 Consider the linear mapping $L : V_n(F) \rightarrow W_m(F)$ and suppose $\{v_1, v_2, \dots, v_n\}$ is a basis for $V_n(F)$, then $\mathcal{R}(L)$ is spanned by the set $\{L(v_1), L(v_2), \dots, L(v_n)\}$ or equivalently $\mathcal{R}(L) = [L(v_1), L(v_2), \dots, L(v_n)]$.

Proof. We must show that $\mathcal{R}(L) \subset [L(v_1), L(v_2), \dots, L(v_n)]$ and $[L(v_1), L(v_2), \dots, L(v_n)] \subset \mathcal{R}(L)$.

i) Showing $\mathcal{R}(L) \subset [L(v_1), L(v_2), \dots, L(v_n)]$: Let $w \in \mathcal{R}(L)$, hence there exists $v \in V_n(F)$ such that $w = L(v)$. Expressing v in the basis $B = \{v_1, v_2, \dots, v_n\}$ gives uniquely $w = L(\sum_{i=1}^n \alpha_i v_i) = \sum_{i=1}^n \alpha_i L(v_i)$, hence $w \in [L(v_1), L(v_2), \dots, L(v_n)]$ implying $\mathcal{R}(L) \subset [L(v_1), L(v_2), \dots, L(v_n)]$

ii) Showing $[L(v_1), L(v_2), \dots, L(v_n)] \subset \mathcal{R}(L)$: Let $w \in [L(v_1), L(v_2), \dots, L(v_n)]$, it follows that $w = \sum_{i=1}^n \alpha_i L(v_i)$ and using linearity $w = L(\sum_{i=1}^n \alpha_i v_i)$ implying that w is the image of the vector $\sum_{i=1}^n \alpha_i v_i$ of $V_n(F)$, hence a vector of $\mathcal{R}(L)$ leading to $[L(v_1), L(v_2), \dots, L(v_n)] \subset \mathcal{R}(L)$. ■

This result states that $\mathcal{R}(L)$ can be generated from the images of the vectors of any basis for the domain $V_n(F)$.

We have seen previously that a span remains unchanged if we remove linearly dependent vectors from the spanning set. Clearly if we remove linearly dependent vectors from $\{L(v_1), L(v_2), \dots, L(v_n)\}$, we will not change $[L(v_1), L(v_2), \dots, L(v_n)]$. It follows that the **remaining vectors** in $\{L(v_1), L(v_2), \dots, L(v_n)\}$ are linearly independent and span $\mathcal{R}(L)$ constituting then a basis for $\mathcal{R}(L)$ by definition.

This suggests an approach to construct a basis for $\mathcal{R}(L)$: Select any basis $\{v_1, v_2, \dots, v_n\}$ for $V_n(F)$, obtain their images $\{L(v_1), L(v_2), \dots, L(v_n)\}$, remove from this set the linearly dependent vectors, then the remaining vectors

constitute a basis. The number of vectors in this basis is by definition the dimension of $\mathcal{R}(L)$, hence the rank of the linear mapping L .

Definition 50 *The null space or Kernel of the mapping $L : V_n(F) \rightarrow W_m(F)$, denoted $\mathcal{N}(L)$, is defined as $\mathcal{N}(L) = \{v \in V_n(F) / L(v) = \theta\}$; The dimension of $\mathcal{N}(L)$ is called the **nullity** of L and denoted $n(L)$.*

$\mathcal{N}(L)$ consists then of the set of all vectors of $V_n(F)$ mapping into the zero vector of $W_m(F)$. This subset of $V_n(F)$ can easily be shown to be closed under linear combinations making of it a subspace of $V_n(F)$.

Example 51 *Consider again the linear mapping $D : P_3 \rightarrow P_3$ defined by $D[p(x)] = \frac{dp(x)}{dx}$ where $P_3 = \{p(x) / p(x) = a_0 + a_1x + a_2x^2 + a_3x^3\}$. It follows that $\mathcal{N}(D) = \{p(x) / D[p(x)] = \theta\}$ where θ is the zero polynomial in this case. $D[p(x)] = \frac{dp(x)}{dx} = \theta$ leads to $\frac{d}{dx}(a_0 + a_1x + a_2x^2 + a_3x^3) = a_1 + 2a_2x + 3a_3x^2 = 0.1 + 0.x + 0.x^2 + 0.x^3$ hence $a_1 = a_2 = a_3 = 0$ and $\mathcal{N}(D) = \{p(x) = a_0\}$ or the set of all zero degree polynomials. The nullity $n(D)$ is 1 since a basis for $\mathcal{N}(D)$ is $\{1\}$.*

From the last two examples we notice that $\dim[\mathcal{N}(D)] + \dim[\mathcal{R}(D)] = \dim[P_3]$. Can this be generalized?

Theorem 52 *Given the linear mapping $L : V_n(F) \rightarrow W_m(F)$, the rank and nullity of L are related by $n(L) + r(L) = n$ i.e. the dimension of the domain.*

Proof. Suppose $n(L) = k$ and let $\{v_1, v_2, \dots, v_k\}$ be a basis for $\mathcal{N}(L)$. Now these vectors constitute a linearly independent subset of vectors of $V_n(F)$ and since any linearly independent subset can be extended to be a basis (shown earlier), we can extend $\{v_1, v_2, \dots, v_k\}$ to be a basis for $V_n(F)$ as $\{v_1, v_2, \dots, v_k, v_{k+1}, \dots, v_n\}$ where $(n - k)$ linearly independent vectors were added. Consider now the set of images of the added vectors $\{L(v_{k+1}), L(v_{k+2}), \dots, L(v_n)\}$ which can be shown to be a basis for $\mathcal{R}(L)$. To be so, the set must be linearly independent and must span $\mathcal{R}(L)$.

i) $\{L(v_{k+1}), L(v_{k+2}), \dots, L(v_n)\}$ linearly independent: let $\sum_{i=k+1}^n \alpha_i L(v_i) = \theta$ and show $\alpha_i = 0$ for $k+1 \leq i \leq n$. By linearity we can write $L(\sum_{i=k+1}^n \alpha_i v_i) = \theta$ meaning that $(\sum_{i=k+1}^n \alpha_i v_i)$ is a vector of $\mathcal{N}(L)$. Thus it can be expressed uniquely using the basis $\{v_1, v_2, \dots, v_k\}$ as $\sum_{i=k+1}^n \alpha_i v_i = \sum_{i=1}^k \alpha_i v_i$ or $\sum_{i=1}^k \alpha_i v_i - \sum_{i=k+1}^n \alpha_i v_i = \theta$ obtaining a linear combination of $\{v_1, v_2, \dots, v_k, v_{k+1}, \dots, v_n\}$ set to zero. Since these vectors are linearly independent by construction, we conclude that $\alpha_i = 0$ for all $1 \leq i \leq n$, thus $\alpha_i = 0$ for $k+1 \leq i \leq n$ and the set $\{L(v_{k+1}), L(v_{k+2}), \dots, L(v_n)\}$ is linearly independent.

ii) $\{L(v_{k+1}), L(v_{k+2}), \dots, L(v_n)\}$ spans $\mathcal{R}(L)$: we must show $[L(v_{k+1}), L(v_{k+2}), \dots, L(v_n)] = \mathcal{R}(L)$ equivalently $[L(v_{k+1}), L(v_{k+2}), \dots, L(v_n)] \subset \mathcal{R}(L)$ and $\mathcal{R}(L) \subset [L(v_{k+1}), L(v_{k+2}), \dots, L(v_n)]$. Now $[L(v_{k+1}), L(v_{k+2}), \dots, L(v_n)] \subset \mathcal{R}(L)$ is obvious since every vector of $[L(v_{k+1}), L(v_{k+2}), \dots, L(v_n)]$ is a linear combination of images hence a vector of $\mathcal{R}(L)$. Let $w \in \mathcal{R}(L)$, it follows that there exists $v \in V_n(F)$ such that $w = L(v)$; Expressing v using the basis $\{v_1, v_2, \dots, v_k, v_{k+1}, \dots, v_n\}$ leads to

$v = \sum_{i=1}^n \alpha_i L(v_i)$, hence $w = L(\sum_{i=1}^n \alpha_i v_i)$ and by linearity $w = \sum_{i=1}^n \alpha_i L(v_i)$ which can be split into $w = \sum_{i=1}^k \alpha_i L(v_i) + \sum_{i=k+1}^n \alpha_i L(v_i)$. Clearly $\sum_{i=1}^k \alpha_i L(v_i) = \theta$ since v_i for $1 \leq i \leq k$ are vectors of $\mathcal{N}(L)$ whose images are zero. It follows that $w = \sum_{i=k+1}^n \alpha_i L(v_i)$ i.e. w is a linear combination of the set $\{L(v_{k+1}), L(v_{k+2}), \dots, L(v_n)\}$, thus $w \in [L(v_{k+1}), L(v_{k+2}), \dots, L(v_n)]$ and $\mathcal{R}(L) \subset [L(v_{k+1}), L(v_{k+2}), \dots, L(v_n)]$. Finally we conclude that $\{L(v_{k+1}), L(v_{k+2}), \dots, L(v_n)\}$ is indeed a basis for $\mathcal{R}(L)$ establishing that $\dim \mathcal{N}(L) + \dim \mathcal{R}(L) = \dim V_n(F) = n$ ■

This result is useful in the sense that the nullity can be determined from the rank and vice-versa.

In this section, we have seen that every linear mapping L defines two fundamental subspaces: $\mathcal{N}(L)$ in the domain and $\mathcal{R}(L)$ in the codomain whose dimensions are related. Furthermore $\mathcal{R}(L)$ is spanned by the images of the vectors of any basis for the domain of L .

Example 53 For the linear mapping $D : P_3 \rightarrow P_3$ defined by $D[p(x)] = \frac{dp(x)}{dx}$ seen earlier, we found $n(D) = 1$ and $r(D) = 3$ leading to $n(D) + r(D) = 4$ which is the dimension of P_3 .

3.4 Nonsingular Linear Mappings

Among linear mappings, there exists a special class of mappings called **non-singular** or **invertible** mappings having the important characteristic that a vector can be identified from its image. In fact, these linear mappings form a subclass of general one-to-one mappings and are considered in this section. Before defining a singular mapping, we need to define a special mapping called the **identity mapping**. Given the vector space $V_n(F)$, we define the identity mapping I as the mapping $I : V_n(F) \rightarrow V_n(F)$ such that $I(v) = v$ for any $v \in V_n(F)$.

Definition 54 A linear mapping $L : V_n(F) \rightarrow W_m(F)$ is said to be nonsingular or invertible if and only if there exists a mapping $L^{-1} : \mathcal{R}(L) \rightarrow V_n(F)$ such that $L^{-1} \circ L = I$ where I is the identity mapping.

Clearly if w is the image of v under the mapping L i.e. $w = L(v)$ and L is nonsingular we have $(L^{-1} \circ L)(v) = L^{-1}[L(v)] = L^{-1}(w) = I(v) = v$ allowing to determine v from its image w . It can easily be shown (do it as an exercise) that L^{-1} is also linear.

This definition tells us what is a nonsingular mapping but it is not very helpful in testing whether or not a given linear mapping is nonsingular. The following result offers several easier ways for this testing.

Theorem 55 Given the linear mapping $L : V_n(F) \rightarrow W_m(F)$, then the following statements are equivalent:

- i) L nonsingular
- ii) L^{-1} exists

- iii) $\mathcal{N}(L) = [\theta]$ or $n(L) = 0$
- iv) $r(L) = n$
- v) L preserves linear independence

Proof. Since all statements are equivalent, we must prove that $i) \Rightarrow ii) \Rightarrow iii) \Rightarrow iv) \Rightarrow v) \Rightarrow i)$

$i) \Rightarrow ii)$: Suppose L nonsingular, then L^{-1} exists by definition.

$ii) \Rightarrow iii)$: Suppose L^{-1} exists and let $v \in \mathcal{N}(L)$ meaning $L(v) = \theta$, then $L^{-1}[L(v)] = L^{-1}(\theta) = \theta$ or $(L^{-1} \circ L)(v) = \theta$ leading to $I(v) = v = \theta$. It follows that any vector of $\mathcal{N}(L)$ is the zero vector θ , hence $\mathcal{N}(L) = [\theta]$ and $\dim \mathcal{N}(L) = \dim[\theta]$ or $n(L) = 0$

$iii) \Rightarrow iv)$: Suppose $n(L) = 0$, then by $n(L) + r(L) = n$ we conclude that $r(L) = 0$

$iv) \Rightarrow v)$: Let $\{v_1, v_2, \dots, v_k\}$ be a set of linearly independent vectors and consider the set $\{L(v_1), L(v_2), \dots, L(v_k)\}$. Setting $\sum_{i=1}^k \alpha_i L(v_i) = \theta$ and using linearity we have $L(\sum_{i=1}^k \alpha_i v_i) = \theta$ meaning that $(\sum_{i=1}^k \alpha_i v_i) \in \mathcal{N}(L)$ and since $\mathcal{N}(L) = [\theta]$, we conclude that $\sum_{i=1}^k \alpha_i v_i = \theta$. Now by assumption $\{v_1, v_2, \dots, v_k\}$ are linearly independent implying that $\alpha_i = 0$ for $1 \leq i \leq k$. It follows that $\{L(v_1), L(v_2), \dots, L(v_k)\}$ are also linearly independent showing that L preserves a nonsingular mapping preserves linear independence.

$v) \Rightarrow i)$: Consider $u, v \in V_n(F)$ such that $u \neq v$ and let $\{v_1, v_2, \dots, v_n\}$ be a basis for $V_n(F)$. We can write uniquely $u = \sum_{i=1}^n \alpha_i v_i$ and $v = \sum_{i=1}^n \beta_i v_i$. Now $L(u) = L(\sum_{i=1}^n \alpha_i v_i)$ and $L(v) = L(\sum_{i=1}^n \beta_i v_i)$ leading to $L(u) - L(v) = L(\sum_{i=1}^n (\alpha_i - \beta_i) v_i)$ or $L(u) - L(v) = \sum_{i=1}^n (\alpha_i - \beta_i) L(v_i)$. Now suppose $L(u) = L(v)$, then $L(u) - L(v) = \sum_{i=1}^n (\alpha_i - \beta_i) L(v_i) = \theta$. Now since $u \neq v$, at least one $(\alpha_i - \beta_i)$ is nonzero meaning that the set $\{L(v_1), L(v_2), \dots, L(v_k)\}$ is linearly dependent contradicting the hypothesis of linear independence preservation. Hence the supposition $L(u) = L(v)$ is necessarily wrong meaning that $u \neq v$ implies $L(u) \neq L(v)$ and the mapping L is one-to-one or nonsingular. ■

Any of the above statements can be used to test the nonsingularity of a given linear mapping. Perhaps the simplest is to set $L(v) = \theta$ and show that the **unique solution** is $v = \theta$ since $\mathcal{N}(L) = [\theta]$ for nonsingular mappings.

If $\{v_1, v_2, \dots, v_n\}$ is a basis for $V_n(F)$, then for a nonsingular mapping $\{L(v_1), L(v_2), \dots, L(v_n)\}$ is a basis for $\mathcal{R}(L)$: this is so because L preserves linear independence and $r(L) = n$.

