

Linear Transformation

Question 01:

Let us consider the basis $S = \{v_1, v_2, v_3\}$ for \mathbb{R}^3 , where $v_1 = (1, 1, 1)$, $v_2 = (1, 1, 0)$ and $v_3 = (1, 0, 0)$.

Let $T : \mathbb{R}^3 \rightarrow \mathbb{R}^2$ be the linear transformation such that $T(v_1) = (1, 0)$, $T(v_2) = (2, -1)$, $T(v_3) = (4, 3)$.

Let (x, y, z) be a vector in \mathbb{R}^3 , Since the set S is basis for \mathbb{R}^3 , then there exists scalars $c_1, c_2, c_3 \in \mathbb{R}$ such that

$$\begin{aligned}c_1 v_1 + c_2 v_2 + c_3 v_3 &= (x, y, z) \\ \Rightarrow c_1(1, 1, 1) + c_2(1, 1, 0) + c_3(1, 0, 0) &= (x, y, z) \\ \Rightarrow (c_1, c_1, c_1) + (c_2, c_2, 0) + (c_3, 0, 0) &= (x, y, z) \\ \Rightarrow c_1 + c_2 + c_3 &= x \\ c_1 + c_2 &= y \\ c_1 = z \Rightarrow c_2 = y - c_1 = y - z \quad \text{and} \quad c_3 = x - (c_1 + c_2) &= x - y.\end{aligned}$$

Therefore $(x, y, z) = z(1, 1, 1) + (y - z)(1, 1, 0) + (x - y)(1, 0, 0)$.

Therefore the linear transformation is given by

$$\begin{aligned}T(x, y, z) &= T\{z(1, 1, 1) + (y - z)(1, 1, 0) + (x - y)(1, 0, 0)\} \\ &= T\{z(1, 1, 1)\} + T\{(y - z)(1, 1, 0)\} + T\{(x - y)(1, 0, 0)\} \{\cdot \cdot T \text{ is linear mapping}\} \\ &= zT(1, 1, 1) + (y - z)T(1, 1, 0) + (x - y)T(1, 0, 0) \{\cdot \cdot T \text{ is linear mapping}\} \\ &= zT(v_1) + (y - z)T(v_2) + (x - y)T(v_3) \\ &= z(1, 0) + (y - z)(2, -1) + (x - y)(4, 3) \\ &= (z, 0) + (2(y - z), -(y - z)) + (4(x - y), 3(x - y)) \\ &= (z, 0) + (2y - 2z, z - y) + (4x - 4y, 3x - 3y) \\ &= (z + 2y - 2z + 4x - 4y, 0 + z - y + 3x - 3y) \\ &= (4x - 2y - z, 3x - 4y + z)\end{aligned}$$

The linear transformation is given by $T(x_1, x_2, x_3) = (4x_1 - 2x_2 - x_3, 3x_1 - 4x_2 + x_3)$.

We know that when T is a linear transformation then $T(xv_1 + yv_2) = xT(v_1) + yT(v_2)$, where $x, y \in \mathbb{R}$ and $v_1, v_2 \in \mathbb{R}^3$.

From the above problem we get $T(x_1, x_2, x_3) = (4x_1 - 2x_2 - x_3, 3x_1 - 4x_2 + x_3)$, $(x_1, x_2, x_3) \in \mathbb{R}^3$.
Then from the above formula we get

$$\begin{aligned}T(2, -3, 5) &= (\{4 \times 2 - 2 \times (-3) - 5\}, \{3 \times 2 - 4 \times (-3) + 5\}) \\ &= (\{8 + 6 - 5\}, \{6 + 12 + 5\}) \\ &= (9, 23).\end{aligned}$$

The final answers are given by

* The linear transformation is given by $T(x_1, x_2, x_3) = (4x_1 - 2x_2 - x_3, 3x_1 - 4x_2 + x_3)$.

* $T(2, -3, 5) = (9, 23)$.

Question 02:

Note that a map $f : A(\neq \phi) \rightarrow B(\neq \phi)$ is linear \Leftrightarrow
 $f(\alpha a + b) = \alpha f(a) + f(b) \forall a, b \in A$ and for any scalar
 $\alpha \in F$ where F is a field.

Here I suppose the field \mathbb{R} .

a) This is linear. Because for any $X = (x, y), Y = (r, z)$ and any $a \in \mathbb{R}$, we have
 $F(aX + Y) = F(ax + r, y + z) = (2(ax + r), y + z)$
 $\Rightarrow F(aX + Y) = (2(ax), y) + (2r, z) = a(2x, y) + (2r, z)$
 $\Rightarrow F(aX + Y) = aF(X) + F(Y)$
 $\Rightarrow F$ is linear.

b) This is not linear. For example,
 $F((1, 1) + (1, 0)) = F(2, 1) = (4, 1)$ but
 $F(1, 1) + F(1, 0) = (1, 1) + (1, 0) = (2, 1) \neq (4, 1).$

c) This is linear. Because for any $\alpha \in \mathbb{R}$,
and for any $A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$ and $B = \begin{bmatrix} e & f \\ g & h \end{bmatrix}$, we have
 $F(\alpha A + B) = F\left(\begin{bmatrix} \alpha a + e & \alpha b + f \\ \alpha c + g & \alpha d + h \end{bmatrix}\right)$
 $= (\alpha b + f) + (\alpha c + g)$
 $= \alpha(b + c) + (f + g)$
 $= \alpha F\left(\begin{bmatrix} a & b \\ c & d \end{bmatrix}\right) + F\left(\begin{bmatrix} e & f \\ g & h \end{bmatrix}\right) = \alpha F(A) + F(B)$
 $\Rightarrow F$ is linear.

d) This is not linear because under linear map, image of zero element is zero element but here

~~$F\left(\begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}\right) = \begin{bmatrix} 0 \\ 0 \\ 3 \end{bmatrix} \neq \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \Rightarrow F \text{ is not linear.}$~~ It is linear, zero maps to zero.

e) This is linear. Because for any $a \in \mathbb{R}$, and any matrices C, D , we have
 $T(aC + D) = (aC + D)B = (aC)B + DB$
 $\Rightarrow T(aC + D) = a(CB) + DB = aT(C) + T(D)$
 $\Rightarrow T$ is linear.

I proved that the maps given in a), c) and e) are LINEAR and the maps given in b) and d) are NOT LINEAR.

Question 03:

Q.3.(a) Let $T : \mathbb{R}^3 \rightarrow \mathbb{R}^2$ be a linear transformation such that $T(e_1) = (1, 1)$, $T(e_2) = (3, 0)$ and $T(e_3) = (4, -7)$, where $\{e_1 = (1, 0, 0), e_2 = (0, 1, 0), e_3 = (0, 0, 1)\}$ be the standard basis for \mathbb{R}^3 .

Let $(1, 3, 8) \in \mathbb{R}^3$, then

$$(1, 3, 8) = (1, 0, 0) + (0, 3, 0) + (0, 0, 8) = (1, 0, 0) + 3(0, 1, 0) + 8(0, 0, 1) = e_1 + 3e_2 + 8e_3.$$

$$\begin{aligned}\therefore T(1, 3, 8) &= T(e_1 + 3e_2 + 8e_3) \\ &= T(e_1) + T(3e_2) + T(8e_3) \{\because T \text{ is a linear transformation}\} \\ &= T(e_1) + 3T(e_2) + 8T(e_3) \{\because T \text{ is a linear transformation}\} \\ &= (1, 1) + 3(3, 0) + 8(4, -7) \\ &= (1, 1) + (9, 0) + (32, -56) \\ &= (1 + 9 + 32, 1 + 0 - 56) \\ &= (42, -55)\end{aligned}$$

(b). Let (x, y, z) be a vector in \mathbb{R}^3 , Since $\{e_1, e_2, e_3\}$ be a standard basis for \mathbb{R}^3 , then there exists scalars c_1, c_2, c_3 such that

$$\begin{aligned}(x, y, z) &= c_1e_1 + c_2e_2 + c_3e_3 \\ &= (c_1, 0, 0) + (0, c_2, 0) + (0, 0, c_3) \\ &= (c_1, c_2, c_3) \\ \Rightarrow c_1 &= x \\ c_2 &= y \\ c_3 &= z\end{aligned}$$

Therefore the linear transformation is given by

$$\begin{aligned}T(x, y, z) &= T(xe_1 + ye_2 + ze_3) \\ &= T(xe_1) + T(ye_2) + T(ze_3) \{\because T \text{ is a linear mapping}\} \\ &= xT(e_1) + yT(e_2) + zT(e_3) \{\because T \text{ is a linear mapping}\} \\ &= x(1, 1) + y(3, 0) + z(4, -7) \\ &= (x, x) + (3y, 0) + (4z, -7z) \\ &= (x + 3y + 4z, x - 7z)\end{aligned}$$

Since \mathbb{R}^2 be a vector space, then

$$c(x, y) = (cx, cy) \quad \text{and} \quad (x, y) + (z, w) = (x + z, y + w) \forall (x, y), (z, w) \in \mathbb{R}^2, c \in \mathbb{R}.$$

(c). The linear transformation is given by $T(x, y, z) = (x + 3y + 4z, x - 7z)$.

Let $\{e_1, e_2, e_3\}$ be the standard basis for \mathbb{R}^3 , and $\{(1, 0), (0, 1)\}$ be the standard basis for \mathbb{R}^2 .

Now

$$\begin{aligned}
 T(e_1) &= T(1, 0, 0) \\
 &= (1 \times 1 + 3 \times 0 + 4 \times 0, 1 - 7 \times 0) \\
 &= (1, 1) = 1(1, 0) + 1(0, 1)
 \end{aligned}$$

$$\begin{aligned}
 T(e_2) &= T(0, 1, 0) \\
 &= (1 \times 0 + 3 \times 1 + 4 \times 0, 0 - 7 \times 0) \\
 &= (0 + 3 + 0, 0 - 0) \\
 &= (3, 0) = 3(1, 0) + 0(0, 1)
 \end{aligned}$$

$$\begin{aligned}
 \text{and } T(e_3) &= T(0, 0, 1) \\
 &= (1 \times 0 + 3 \times 0 + 4 \times 1, 0 - 7 \times 1) \\
 &= (0 + 0 + 4, 0 - 7) \\
 &= (4, -7) = 4(1, 0) - 7(0, 1)
 \end{aligned}$$

The matrix of the linear transformation is $A = \begin{bmatrix} 1 & 3 & 4 \\ 1 & 0 & -7 \end{bmatrix}$.

The final answers are given by

(a). $T(1, 3, 8) = (42, -55)$.

(b). $T(x, y, z) = (x + 3y + 4z, x - 7z)$.

(c). The matrix of the linear transformation is $A = \begin{bmatrix} 1 & 3 & 4 \\ 1 & 0 & -7 \end{bmatrix}$.

Question 06:

Part(a)

DEFINITION 11.7. The **kernel** of a linear transformation $\mathbf{T} : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is the set of all $\mathbf{x} \in \mathbb{R}^n$ such that $\mathbf{T}\mathbf{x} = \mathbf{0} \in \mathbb{R}^m$.

$$\ker(T) = \{\mathbf{x} \in \mathbb{R}^n \mid T(\mathbf{x}) = \mathbf{0}\}$$

LEMMA 11.8. The kernel of a linear transformation $\mathbf{T} : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is a subspace of \mathbb{R}^n .

Proof. $\ker(T)$ is obviously a subset of \mathbb{R}^n . We need to show that it's closed under scalar multiplication and vector addition. Let $\lambda \in \mathbb{R}$ and $\mathbf{x} \in \ker(T)$ be arbitrary elements of their respective sets. Then $T(\lambda\mathbf{x}) = \lambda T(\mathbf{x})$, since T is a linear transformation. But $T(\mathbf{x}) = \mathbf{0}$ since $\mathbf{x} \in \ker(T)$. So $T(\lambda\mathbf{x}) = \mathbf{0}$. We conclude that if $\lambda \in \mathbb{R}$, and $\mathbf{x} \in \ker(T)$, then $\lambda\mathbf{x} \in \ker(T)$, and so $\ker(T)$ is closed under scalar multiplication.

Now let $\mathbf{x}_1, \mathbf{x}_2$ be arbitrary vectors in $\ker(T)$. Then since T is a linear transformation, $T(\mathbf{x}_1 + \mathbf{x}_2) = T(\mathbf{x}_1) + T(\mathbf{x}_2) = \mathbf{0} + \mathbf{0} = \mathbf{0}$ and so $\mathbf{x}_1 + \mathbf{x}_2 \in \ker(T)$. Thus, $\ker(T)$ is closed under vector addition.

Since $\ker(T)$ is a subset of \mathbb{R}^n that is closed under both scalar multiplication and vector addition, it is a subspace of \mathbb{R}^n . \square

DEFINITION 11.9. The **image** or **range** of \mathbf{T} is the set of all $\mathbf{y} \in \mathbb{R}^m$ such that $\mathbf{y} = \mathbf{T}(\mathbf{x})$ for some $\mathbf{x} \in \mathbb{R}^n$.

$$\text{range}(T) = \{\mathbf{y} \in \mathbb{R}^m \mid \mathbf{y} = T(\mathbf{x}) \text{ for some } \mathbf{x} \in \mathbb{R}^n\}$$

LEMMA 11.10. The range of a linear transformation $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is a subspace of \mathbb{R}^m .

Proof. We need to show that $\text{range}(T)$ is closed under both scalar multiplication and vector addition.

Suppose $\mathbf{y} \in \text{range}(T)$. Then there must be an $\mathbf{x} \in \mathbb{R}^n$ such that $\mathbf{y} = T(\mathbf{x})$. But then $\lambda\mathbf{x} \in \mathbb{R}^n$ and

$$T(\lambda\mathbf{x}) = \lambda T(\mathbf{x}) = \lambda\mathbf{y}$$

and so $\lambda\mathbf{y}$ is in $\text{range}(T)$. Hence, $\text{range}(T)$ is closed under scalar multiplication.

Suppose $\mathbf{y}_1, \mathbf{y}_2 \in \text{range}(T)$. Then there must be vectors $\mathbf{x}_1, \mathbf{x}_2 \in \mathbb{R}^n$ such that $\mathbf{y}_1 = T(\mathbf{x}_1)$ and $\mathbf{y}_2 = T(\mathbf{x}_2)$. Now apply T to the vector sum $\mathbf{x}_1 + \mathbf{x}_2$:

$$T(\mathbf{x}_1 + \mathbf{x}_2) = T(\mathbf{x}_1) + T(\mathbf{x}_2) = \mathbf{y}_1 + \mathbf{y}_2$$

This displays $\mathbf{y}_1 + \mathbf{y}_2$ as an element of $\text{range}(T)$.

Since $\text{range}(T) \subset \mathbb{R}^m$ is closed under both scalar multiplication and vector addition, it is a subspace of \mathbb{R}^m . \square

Now let \mathbf{A} be the $m \times n$ matrix corresponding to a linear transformation $\mathbf{T} : \mathbb{R}^n \rightarrow \mathbb{R}^m$. Then

$$\begin{aligned}\ker(\mathbf{T}) &= \{\mathbf{x} \in \mathbb{R}^n \mid \mathbf{T}(\mathbf{x}) = \mathbf{0}\} \\ &= \{\mathbf{x} \in \mathbb{R}^n \mid \mathbf{A}_T \mathbf{x} = \mathbf{0}\} = \text{Null space of } \mathbf{A}_T\end{aligned}$$

$$\begin{aligned}\text{range}(\mathbf{T}) &= \{\mathbf{y} \in \mathbb{R}^m \mid \mathbf{y} = \mathbf{T}(\mathbf{x}) \text{ , for some } \mathbf{x} \in \mathbb{R}^n\} \\ &= \{\mathbf{y} \in \mathbb{R}^m \mid \mathbf{y} = \mathbf{A}_T \mathbf{x} \text{ , for some } \mathbf{x} \in \mathbb{R}^n\} = \text{column space of } \mathbf{A}_T\end{aligned}$$

Part(b)

DEFINITION

If $T: V \longrightarrow W$ is a linear transformation, then the dimension of the range of T is called the **rank of T** and is denoted by $\text{rank}(T)$; the dimension of the kernel is called the **nullity of T** and is denoted by $\text{nullity}(T)$.

If A is an $m \times n$ matrix and $T_A: R^n \longrightarrow R^m$ is multiplication by A , then we know from Example 1 that the kernel of T_A is the nullspace of A and the range of T_A is the column space of A . Thus we have the following relationship between the rank and nullity of a matrix and the rank and nullity of the corresponding matrix transformation.

Part(c)

If A is the matrix representaiton of a linear transformation T , then

1. $\mathcal{N}(T) = \mathcal{N}(A)$ and $\mathcal{R}(T) = \mathcal{R}(A)$.
2. The nullity of T is the same as the nullity of A .
3. The rank of T is the same as the rank of A .

Part(d)

for finding out the kernel we have to do

$$T(\mathbf{x}_1, \mathbf{x}_2, \mathbf{x}_3) = (0, 0, 0)$$

$$\text{so } \mathbf{x}_1 + 2\mathbf{x}_3 + \mathbf{x}_2 = 0, \mathbf{x}_1 + \mathbf{x}_3 = 0, 2\mathbf{x}_1 + \mathbf{x}_2 + 3\mathbf{x}_3 = 0$$

$$\text{so } \mathbf{x}_1 = -\mathbf{x}_3$$

$$\text{implies also we get, } \mathbf{x}_2 + \mathbf{x}_3 = 0$$

$$\text{so } (\mathbf{x}_1, \mathbf{x}_2, \mathbf{x}_3) = (-\mathbf{x}_3, -\mathbf{x}_3, \mathbf{x}_3)$$

$$\text{so dimension of kernel is 1 and bases is } \{(-1, -1, 1)\}$$

and for range we know we have to collect linearly independent coulms

$$\text{here columns are } \begin{bmatrix} 1 \\ 1 \\ 2 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}, \begin{bmatrix} 2 \\ 1 \\ 3 \end{bmatrix}$$

$$\text{so } \begin{bmatrix} 1 \\ 1 \\ 2 \end{bmatrix} + \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 2 \\ 1 \\ 3 \end{bmatrix}$$

so linearly independent columns is first and second column

$$\text{so basis for range is } \left\{ \begin{bmatrix} 1 \\ 1 \\ 2 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} \right\}$$

dimesion of range space is 2

Question 08:

Let A denote an $m \times n$ matrix of rank r and let $T_A : \mathbb{R}^n \rightarrow \mathbb{R}^m$ denote the corresponding matrix transformation given by $T_A(\mathbf{x}) = A\mathbf{x}$ for all columns \mathbf{x} in \mathbb{R}^n . It follows from Example 7.2.1 and Example 7.2.2 that $\text{im } T_A = \text{col } A$, so $\dim(\text{im } T_A) = \dim(\text{col } A) = r$. On the other hand Theorem 5.4.2 shows that $\dim(\ker T_A) = \dim(\text{null } A) = n - r$. Combining these we see that

$$\dim(\text{im } T_A) + \dim(\ker T_A) = n \quad \text{for every } m \times n \text{ matrix } A$$

The main result of this section is a deep generalization of this observation.

Theorem 7.2.4: Dimension Theorem

Let $T : V \rightarrow W$ be any linear transformation and assume that $\ker T$ and $\text{im } T$ are both finite dimensional. Then V is also finite dimensional and

$$\dim V = \dim(\ker T) + \dim(\text{im } T)$$

In other words, $\dim V = \text{nullity}(T) + \text{rank}(T)$.

Proof. Every vector in $\text{im } T = T(V)$ has the form $T(\mathbf{v})$ for some \mathbf{v} in V . Hence let $\{T(\mathbf{e}_1), T(\mathbf{e}_2), \dots, T(\mathbf{e}_r)\}$ be a basis of $\text{im } T$, where the \mathbf{e}_i lie in V . Let $\{\mathbf{f}_1, \mathbf{f}_2, \dots, \mathbf{f}_k\}$ be any basis of $\ker T$. Then $\dim(\text{im } T) = r$ and $\dim(\ker T) = k$, so it suffices to show that $B = \{\mathbf{e}_1, \dots, \mathbf{e}_r, \mathbf{f}_1, \dots, \mathbf{f}_k\}$ is a basis of V .

1. B spans V . If \mathbf{v} lies in V , then $T(\mathbf{v})$ lies in $\text{im } T$, so

$$T(\mathbf{v}) = t_1 T(\mathbf{e}_1) + t_2 T(\mathbf{e}_2) + \dots + t_r T(\mathbf{e}_r) \quad t_i \text{ in } \mathbb{R}$$

This implies that $\mathbf{v} - t_1 \mathbf{e}_1 - t_2 \mathbf{e}_2 - \dots - t_r \mathbf{e}_r$ lies in $\ker T$ and so is a linear combination of $\mathbf{f}_1, \dots, \mathbf{f}_k$. Hence \mathbf{v} is a linear combination of the vectors in B .

2. B is linearly independent. Suppose that t_i and s_j in \mathbb{R} satisfy

$$t_1 \mathbf{e}_1 + \dots + t_r \mathbf{e}_r + s_1 \mathbf{f}_1 + \dots + s_k \mathbf{f}_k = \mathbf{0} \quad (7.1)$$

Applying T gives $t_1 T(\mathbf{e}_1) + \dots + t_r T(\mathbf{e}_r) = \mathbf{0}$ (because $T(\mathbf{f}_i) = \mathbf{0}$ for each i). Hence the independence of $\{T(\mathbf{e}_1), \dots, T(\mathbf{e}_r)\}$ yields $t_1 = \dots = t_r = 0$. But then (7.1) becomes

$$s_1 \mathbf{f}_1 + \dots + s_k \mathbf{f}_k = \mathbf{0}$$

so $s_1 = \dots = s_k = 0$ by the independence of $\{\mathbf{f}_1, \dots, \mathbf{f}_k\}$. This proves that B is linearly independent. \square

Note that the vector space V is not assumed to be finite dimensional in Theorem 7.2.4. In fact, verifying that $\ker T$ and $\text{im } T$ are both finite dimensional is often an important way to *prove* that V is finite dimensional.

Note further that $r + k = n$ in the proof so, after relabelling, we end up with a basis

$$B = \{\mathbf{e}_1, \mathbf{e}_2, \dots, \mathbf{e}_r, \mathbf{e}_{r+1}, \dots, \mathbf{e}_n\}$$

of V with the property that $\{\mathbf{e}_{r+1}, \dots, \mathbf{e}_n\}$ is a basis of $\ker T$ and $\{T(\mathbf{e}_1), \dots, T(\mathbf{e}_r)\}$ is a basis of $\text{im } T$. In fact, if V is known in advance to be finite dimensional, then *any* basis $\{\mathbf{e}_{r+1}, \dots, \mathbf{e}_n\}$ of $\ker T$ can be extended to a basis $\{\mathbf{e}_1, \mathbf{e}_2, \dots, \mathbf{e}_r, \mathbf{e}_{r+1}, \dots, \mathbf{e}_n\}$ of V by Theorem 6.4.1. Moreover, it turns out that, no matter how this is done, the vectors $\{T(\mathbf{e}_1), \dots, T(\mathbf{e}_r)\}$ will be a basis of $\text{im } T$. This result is useful, and we record it for reference. The proof is much like that of Theorem 7.2.4 and is left as Exercise 7.2.26.