Linear Algebra

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Outline

- 1. Definition.
- 2. Matrix transformation.
- 3. Matrix transformation and bases.
- 4. Kernal of the transformation (Null space).

1. Definition

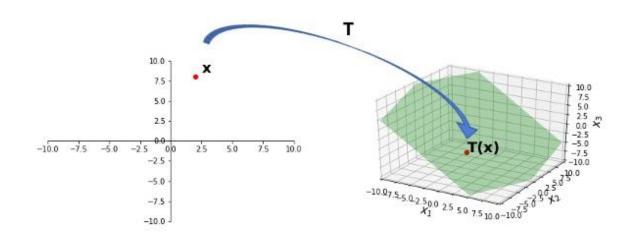
Introduction to Linear Transformations

 \blacksquare Function T that maps a vector space V into a vector space W:

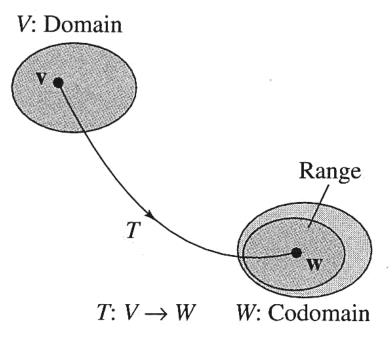
 $T: V \xrightarrow{mapping} W$,

V: the domain of T

W: the codomain of T



V, W: vector space



Definitions

■ Image of **v** under *T*:

If v is in V and w is in W such that

$$T(\mathbf{v}) = \mathbf{w}$$

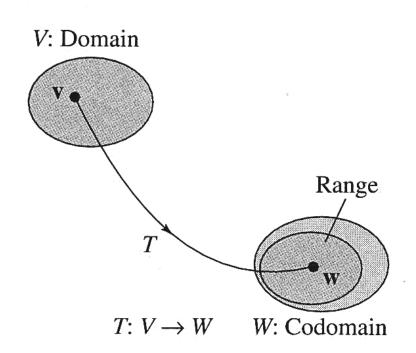
Then \mathbf{w} is called the image of \mathbf{v} under T.

• the range of T:

The set of all images of vectors in V.

• the preimage of w:

The set of all \mathbf{v} in V such that $T(\mathbf{v}) = \mathbf{w}$.



Ex: (A function from R^2 into R^2)

$$T: R^{2} \to R^{2} \quad \mathbf{v} = (v_{1}, v_{2}) \in R^{2}$$

$$T(v_{1}, v_{2}) = (v_{1} - v_{2}, v_{1} + 2v_{2})$$

$$T(v_{1}, v_{2}) = (v_{1} - v_{2}, v_{1} + 2v_{2})$$

$$T(v_{2}, v_{2}) = (v_{1} - v_{2}, v_{1} + 2v_{2})$$

(a) Find the image of $\mathbf{v}=(-1,2)$. (b) Find the preimage of $\mathbf{w}=(-1,11)$

Sol:
$$v_1 = v_1$$
 $v_2 = v_2$ $v_3 = v_2 = v_2$ $v_1 = v_2 = v_1 + 2v_2 = 11$ $v_1 = v_2 = 11$ $v_2 = v_3 = v_4 = v_2 = 11$ $v_1 = v_2 = v_3 = v_4 = v_4$

■Linear Transformation (L.T.):

V,W: vector space

 $T:V \to W: V$ to W linear transformation

(1)
$$T(\mathbf{u} + \mathbf{v}) = T(\mathbf{u}) + T(\mathbf{v}), \forall \mathbf{u}, \mathbf{v} \in V$$

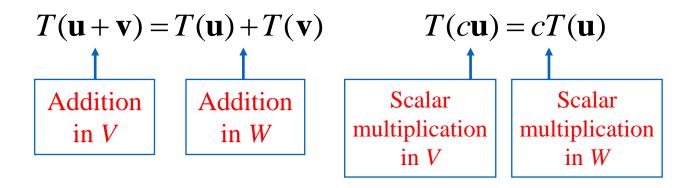
(2)
$$T(c\mathbf{u}) = cT(\mathbf{u}), \forall c \in R$$

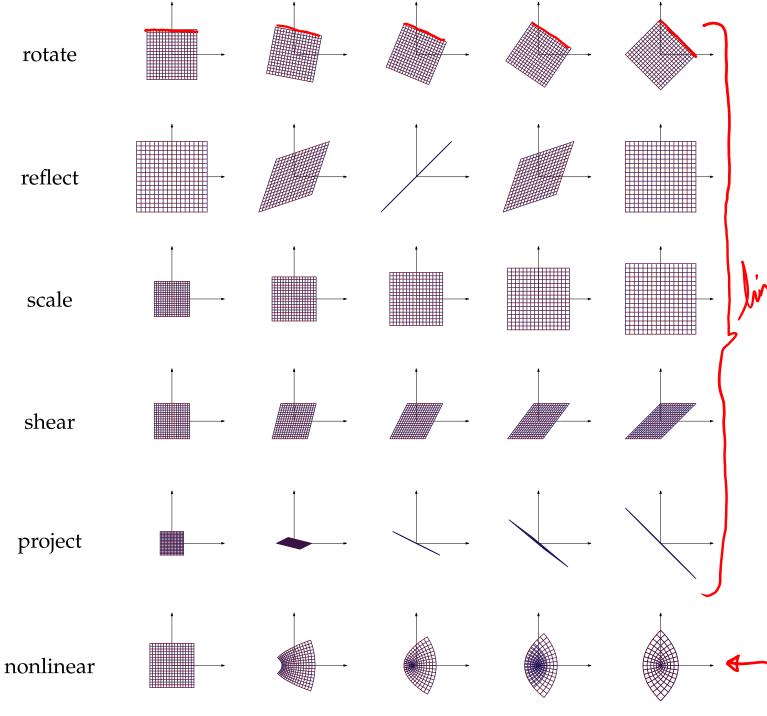
Or in one condition

$$T(\alpha \mathbf{u} + \beta \mathbf{v}) = \alpha T(\mathbf{u}) + \beta T(\mathbf{v})$$

Note:

A linear transformation is said to be operation preserving.





- Many fundamental geometric transformations are linear.
- The figure below illustrates several linear transformations (as well as one nonlinear one, for comparison) from the plane to the plane.
- The leftmost column shows a square grid of points, and the rightmost column shows the images of those points.
- The other columns show each point somewhere along the path from its original location in the domain to its final location in the codomain, to help you get a sense of which points go where.

Verifying a linear transformation

Ex: (Verifying a linear transformation
$$T$$
 from R^2 into R^2)
$$T(\overset{\times}{v_1},\overset{\times}{v_2}) = (\overset{\times}{v_1} - \overset{\times}{v_2}, \overset{\times}{v_1} + 2v_2) \quad \text{fig. d. transformation} \quad (\overset{\times}{v_1},\overset{\times}{v_2}) + (\overset{\times}{p}v_1,\overset{\times}{p}v_2)$$
Pf:
$$\mathbf{u} = (\overset{\times}{u_1},\overset{\times}{u_2}), \quad \mathbf{v} = (v_1,v_2) \quad \text{vector in } R^2, \quad \alpha,\beta: \text{ any real number}$$

We need to prove that
$$T(\alpha \mathbf{u} + \beta \mathbf{v}) = \alpha T(\mathbf{u}) + \beta T(\mathbf{v})$$

 $LHS = T(\alpha \mathbf{u} + \beta \mathbf{v}) = T(\alpha u_1 + \beta v_1, \alpha u_2 + \beta v_2)$
 $= ((\alpha u_1 + \beta v_1) - (\alpha u_2 + \beta v_2), (\alpha u_1 + \beta v_1) + 2(\alpha u_2 + \beta v_2))$
 $= (\alpha u_1 + \beta v_1 - \alpha u_2 - \beta v_2, \alpha u_1 + \beta v_1 + 2\alpha u_2 + 2\beta v_2)$
 $RHS = \alpha T(\mathbf{u}) + \beta T(\mathbf{v}) = \alpha (u_1 - u_2, u_1 + 2u_2) + \beta (v_1 - v_2, v_1 + 2v_2)$
 $= (\alpha u_1 - \alpha u_2 + \beta v_1 - \beta v_2, \alpha u_1 + 2\alpha u_2 + \beta v_1 + 2\beta v_2) = LHS$

$$= (\alpha u_1 + \beta v_1 - \alpha u_2 - \beta v_2, \alpha u_1 + \beta v_1 + 2 \alpha u_2 + 2 \beta v_2)$$

$$RHS = \alpha T(\mathbf{u}) + \beta T(\mathbf{v}) = \alpha (u_1 - u_2, u_1 + 2u_2) + \beta (v_1 - v_2, v_1 + 2v_2)$$

$$= (\alpha u_1 - \alpha u_2 + \beta v_1 - \beta v_2), \alpha u_1 + 2 \alpha u_2 + \beta v_1 + 2 \beta v_2) = LHS$$

Therefore, T is a linear transformation.

Verifying a linear transformation

Ex: Let
$$L: R^3 \to R^3$$
, $L(\mathbf{x}, \mathbf{y}, \mathbf{z}) = (\mathbf{x} + \mathbf{y}, \mathbf{y}, \mathbf{x} - \mathbf{z})$ for f then.

Is L a linear transformation?

Pf:

Let $\mathbf{u}, \mathbf{v} \in R^3$, $\mathbf{u} = (u_1, u_2, u_3)$, $\mathbf{v} = (v_1, v_2, v_3)$ α, β : any real number.

We need to prove that $L(\alpha \mathbf{u} + \beta \mathbf{v}) = \alpha L(\mathbf{u}) + \beta L(\mathbf{v})$
 $LHS = L(\alpha \mathbf{u} + \beta \mathbf{v}) = L(\alpha u_1 + \beta v_1, \alpha u_2 + \beta v_2, \alpha u_3 + \beta v_3)$
 $\mathbf{u} = (\alpha u_1 + \beta v_1 + \alpha u_2 + \beta v_2, \alpha u_2 + \beta v_2, \alpha u_1 + \beta v_1 - \alpha u_3 - \beta v_3)$
 $RHS = \alpha L(\mathbf{u}) + \beta L(\mathbf{v}) = \alpha (u_1 + u_2, u_2, u_1 - u_3) + \beta (v_1 + v_2, v_2, v_1 - v_3)$
 $\mathbf{u} = (\alpha u_1 + \alpha u_2 + \beta v_1 + \beta v_2, \alpha u_2 + \beta v_2, \alpha u_1 - \alpha u_3 + \beta v_1 - \beta v_3) = LHS$

Therefore, T is a linear transformation.

Verifying a linear transformation

$$\blacksquare \text{Ex: Let } L: R^2 \to R^3, \ L(x,y) = (x+1,y,x+y)$$

Is L a linear transformation?

Pf:

Let
$$u$$
, $v \in R^2$, $u = (u_1, u_2)$, $v = (v_1, v_2)$ α , β : any real number

We need to prove that
$$L(\alpha \mathbf{u} + \beta \mathbf{v}) = \alpha L(\mathbf{u}) + \beta L(\mathbf{v})$$

$$LHS = L(\alpha \mathbf{u} + \beta \mathbf{v}) = L(\alpha u_1 + \beta v_1, \alpha u_2 + \beta v_2)$$

$$= (\alpha u_1 + \beta v_1 + 1, \alpha u_2 + \beta v_2, \alpha u_1 + \beta v_1 + \alpha u_2 + \beta v_2)$$

$$RHS = \alpha L(\mathbf{u}) + \beta L(\mathbf{v}) = \alpha (u_1 + 1, u_2, u_1 + u_2) + \beta (v_1 + 1, v_2, v_1 + v_2)$$

$$= (\alpha u_1 + \alpha + \beta v_1 + \beta \alpha u_2 + \beta v_2, \alpha u_1 + \alpha u_2 + \beta v_1 + \beta v_2) \neq LHS$$

$$=(lpha\ u_1+lpha\ +eta\ v_1+eta\ lpha\ u_2+eta\ v_2$$
 , $lpha\ u_1+lpha\ u_2+eta\ v_1+eta\ v_2)
eq LHS$

Therefore, T is **not** a linear transformation.

2. Matrix transformation

Matrix transformation

$$T(x) = A x$$

Every matrix transformation $T: x \to Ax$ is a linear transformation.

$$A_{m \times n} x_{n \times 1} = b_{m \times 1}$$
 transforms x (in \mathbb{R}^n) into b (in \mathbb{R}^m)

Proof that Ax is a linear transformation:

Let
$$T(x_1) = Ax_1$$
 and $T(x_2) = Ax_2$

We need to prove that

$$T(\alpha x_1 + \beta x_2) = \alpha T(x_1) + \beta T(x_2)$$

 α , β : any real number

$$|LHS = T(\alpha x_1 + \beta x_2) = A(\alpha x_1 + \beta x_2)$$

$$= \alpha A x_1 + \beta A x_2$$

$$= \alpha T(x_1) + \beta T(x_2) = RHS$$

Ex: (A linear transformation defined by a matrix)

The function
$$T: \mathbb{R}^2 \to \mathbb{R}^3$$
 is defined as $T(\mathbf{v}) = A\mathbf{v} = \begin{bmatrix} 3 & 0 \\ 2 & 1 \\ -1 & -2 \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix}$
Find $T(\mathbf{v})$, where $\mathbf{v} = (2, -1)$

Sol: $\mathbf{v} = (2, -1)$

$$T(\mathbf{v}) = A\mathbf{v} = \begin{bmatrix} 3 & 0 \\ 2 & 1 \\ -1 & -2 \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix}$$

$$T(2,-1) = (6,3,0)$$

Ex: Let
$$A = \begin{bmatrix} -1 & 2 & 1 & 3 & 4 \\ 0 & 0 & 2 & -1 & 0 \end{bmatrix}$$
 and let $T: R^5 \to R^2$ be a linear transformation such that $T(X) = AX$

Find (a)
$$T(1, 0, -1, 3, 0)$$
 (b) $T^{-1}(-1, 8)$

(b)
$$T^{-1}(-1,8)$$

(a)
$$T(1,0,-1,3,0) = \begin{bmatrix} -1 & 2 & 1 & 3 & 4 \\ 0 & 0 & 2 & -1 & 0 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \\ -1 \\ 3 \\ 0 \end{bmatrix} = \begin{bmatrix} -1+0-1+9+0 \\ 0+0-2-3+0 \end{bmatrix} = \begin{bmatrix} 7 \\ -5 \end{bmatrix}$$

(b) Let $(u_1, u_2, u_3, u_4, u_5)$ be the pre-image of (-1, 8) under T: T(X) = AX

$$\begin{bmatrix} -1 & 2 & 1 & 3 & 4 \\ 0 & 0 & 2 & -1 & 0 \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \\ u_3 \\ u_4 \\ u_5 \end{bmatrix} = \begin{bmatrix} -1 \\ 8 \end{bmatrix}$$
 System of linear equations
$$\begin{bmatrix} -1 & 2 & 1 & 3 & 4 & -1 \\ 0 & 0 & 2 & -1 & 0 & 8 \end{bmatrix}$$

The augmented matrix

$$\begin{bmatrix} -1 & 2 & 1 & 3 & 4 & -1 \\ 0 & 0 & 2 & -1 & 0 & 8 \end{bmatrix}$$

Ex:(Continued) Let
$$A = \begin{bmatrix} -1 & 2 & 1 & 3 & 4 \\ 0 & 0 & 2 & -1 & 0 \end{bmatrix}$$
 and let

 $T: \mathbb{R}^5 \to \mathbb{R}^2$ be a linear transformation such that T(X) = AX

Find (a)
$$T(1, 0, -1, 3, 0)$$
 (b) $T^{-1}(-1, 8)$

(b)
$$T^{-1}(-1,8)$$

Let
$$U_2 = t_1$$
, $U_4 = t_2$, $U_5 = t_3$
 E_2 : $2U_3 - t_2 = 8 \rightarrow U_3 = 4 + \frac{1}{2}t_2$
 E_1 : $-U_1 + 2t_1 + 4 + \frac{1}{2}t_2 + 3t_2 + 4t_3 = -1$
 $U_1 = 2t_1 + \frac{\pi}{2}t_2 + 4t_3 + 5$
 $U_1 = 2t_1 + \frac{\pi}{2}t_2 + 4t_3 + 5$
 $U_2 = t_1$
 $U_3 = t_2$
 $U_3 = t_2$
 $U_3 = t_3$
 $U_4 = t_2$
 $U_5 = t_3$
 $U_5 = t_3$
 $U_7 = t_7$
 $U_7 = t_7$

3. Matrix transformation and bases

• Ex: (Linear transformations and bases)

Let $T: R^3 \to R^3$ be a linear transformation such that

$$\begin{cases}
T(1,0,0) \neq (2,-1,4) \\
T(0,1,0) \neq (1,5,-2)
\end{cases}$$

$$T(0,0,1) \neq (0,3,1)$$

Linear combination in terms of the basis

$$(2,3,-2) = 2(1,0,0) + 3(0,1,0) - 2(0,0,1)$$

$$T(2,3,-2) = 2T(1,0,0) + 3T(0,1,0) - 2T(0,0,1)$$
 (T is a L.T.)
$$= 2(2,-1,4) + 3(1,5,-2) - 2(0,3,1)$$

$$= (7,7,0)$$

the standard basis
$$\begin{bmatrix}
1 & 0 \\
5 & 3 \\
-2 & 1
\end{bmatrix}
\begin{bmatrix}
2 \\
3 \\
-2
\end{bmatrix}
=
\begin{bmatrix}
7 \\
7 \\
0
\end{bmatrix}$$

$$\begin{bmatrix} 1 & 0 & 0 & 2 \\ 0 & 1 & 0 & 3 \\ 0 & 0 & 1 & -2 \end{bmatrix}$$

$$\begin{bmatrix} V_1 & V_2 & V_3 & \text{Tost} \\ \text{Std. basis} & \text{Std. basis} \end{bmatrix}$$

Matrix transformation

To represent a linear transformation T by a matrix A

- 1. Apply the **transformation** to each vector in the **standard bases** of the domain.
- 2. Place the resulting vectors in the columns of A

$$A = [T(e_{1}), T(e_{2}), T(e_{3}), ..., T(e_{n})]$$

$$\begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix} \begin{bmatrix} 0 \\ 1 \\ 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix} \begin{bmatrix} 0 \\ 0 \\ 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

Ex: Find the matrix A of the following transformation such that T(x) = A x

$$T\left(\begin{bmatrix} x \\ y \end{bmatrix}\right) = \begin{bmatrix} x \\ y \\ x + y \end{bmatrix} \quad \text{for } x \in \mathbb{R}$$

Transformation is from R^2 to $R^3 o Let A = [T(e_1), T(e_2)]$

$$T(e_1) = T\left(\begin{bmatrix}1\\0\end{bmatrix}\right) = \begin{bmatrix}1\\0\\1\end{bmatrix}$$

$$T(e_2) = T\left(\begin{bmatrix}0\\1\end{bmatrix}\right) = \begin{bmatrix}0\\1\\1\end{bmatrix}$$

$$A = \begin{bmatrix}1&0\\0&1\\1&1\end{bmatrix}\begin{bmatrix}x\\y\end{bmatrix} = \begin{bmatrix}x\\y\\x+y\end{bmatrix}$$

• Ex: (Linear transformations and bases)

Standard Basis

Let $T: \mathbb{R}^3 \to \mathbb{R}^2$ be a linear transformation such that

$$T(1,0,0) = (2,-4)$$

$$T(0,1,0) = (3,-5)$$

$$T(0,0,1) \neq (2,3)$$

Find (a)
$$T(1, -2, 3)$$

$$A = \begin{bmatrix} 2 & 3 & 2 \end{bmatrix}$$

$$\begin{bmatrix} 2 & 3 & 2 \\ -4 & -5 & 3 \end{bmatrix} \begin{bmatrix} 1 \\ -2 \\ 3 \end{bmatrix} = \begin{bmatrix} 2 \\ 15 \end{bmatrix}$$

$$T(1,-2,3) = (2,15)$$

(b)
$$T(a,b,c)$$
 for f trans.

$$\begin{bmatrix} 2 & 3 & 2 \\ -4 & -5 & 3 \end{bmatrix} \begin{bmatrix} a \\ b \\ c \end{bmatrix} = \begin{bmatrix} 2a+3b+2c \\ -4a-5b+3c \end{bmatrix}$$

$$= (2 a + 3 b + 2 c, -4 a - 5 b + 3 c)$$

Ex: Let $L: \mathbb{R}^3 \to \mathbb{R}^2$ be a linear transformation for which

$$\overline{L(1,1,0)} = (2,1) , L(1,0,1) = (1,-1) , L(0,1,1) = (0,0)$$
 Non standard Basis Find $L(2,-4,1)$

First, write (2,-4,1) as a linear combination of (1,1,0), (1,0,1), and (0,1,1)

$$\rightarrow$$
 (2, -4, 1) = α (1, 1, 0) + β (1, 0, 1) + γ (0, 1, 1)

From row 3 \rightarrow 2 $\gamma = -5 \rightarrow \gamma = -2.5$

From **row 2**
$$\rightarrow$$
 $-\beta$ + (-2.5) = -6 \rightarrow β = 6 - 2.5 \rightarrow β = 3.5

From **row 1**
$$\rightarrow \alpha$$
 + (3.5) + (0) = 2 $\rightarrow \alpha$ = 2 - 3.5 $\rightarrow \alpha$ = -1.5

Ex:(Continued) Let $L: \mathbb{R}^3 \to \mathbb{R}^2$ be a linear transformation for which

$$L(1,1,0)=(2,1)$$
 , $L(1,0,1)=(1,-1)$, $L(0,1,1)=(0,0)$ Find $L(2,-4,1)$

First, write (2,-4,1) as a linear combination of (1,1,0), (1,0,1), and (0,1,1)

$$\rightarrow$$
 (2, -4, 1) = -1.5 (1, 1, 0) + 3.5 (1, 0, 1) - 2.5 (0, 1, 1)

$$\Rightarrow (2, -4, 1) = -1.5(1, 1, 0) + 3.5(1, 0, 1) - 2.5(0, 1, 1)$$

$$\Rightarrow L(2, -4, 1) = -1.5L(1, 1, 0) + 3.5L(1, 0, 1) - 2.5L(0, 1, 1)$$

$$\rightarrow L(2,-4,1) = -1.5(2,1) + 3.5(1,-1) - 2.5(0,0)$$

$$\rightarrow L(2,-4,1) = (0.5,-5)$$

Ex: Let $L: P_2 \to P_2$ be a linear transformation for which

$$L(x+1)=x$$
 , $L(x-1)=1$, $L(x^2)=0$ Non standard Basis

Find
$$L(-x^2+3x+2)$$

First, write $(-x^2 + 3x + 2)$ as a linear combination of (x + 1), (x - 1), and (x^2) $(-x^2 + 3x + 2) = \alpha (x + 1) + \beta (x - 1) + \gamma (x^2)$

$$(-x^{2} + 3x + 2) = \alpha(x + 1) + \beta(x - 1) + \gamma(x^{2})$$

$$\rightarrow (-x^2 + 3x + 2) = [\alpha - \beta] + [\alpha + \beta]x + [\gamma]x^2$$

$$\rightarrow \alpha - \beta = 2$$

$$\Rightarrow \alpha - \beta = 2$$

$$\Rightarrow \alpha + \beta = 3$$

$$\Rightarrow \beta = 3$$

$$\Rightarrow \alpha = 2$$

$$\Rightarrow \beta = 3$$

$$\Rightarrow \alpha = 2.5$$

$$\Rightarrow \alpha = 2.5$$

$$\Rightarrow \beta = 3 - 2.5 = 0.5$$

$$\rightarrow \gamma = -1$$

$$\rightarrow$$
 Solve

$$2 \alpha = 5 \rightarrow \alpha = 2.5$$

$$\beta = 3 - 2.5 = 0.5$$

$$\begin{bmatrix} 1 & -1 & 0 & 2 \\ 1 & 1 & 0 & 3 \\ 0 & 0 & 1 & -1 \end{bmatrix}$$

$$-R_1 + R_2 \to R_2$$

$$\sim \begin{bmatrix} 1 & -1 & 0 & 2 \\ 0 & 2 & 0 & 1 \\ 0 & 0 & 1 & 1 \end{bmatrix}$$

Ex:(Continued) Let $L: P_2 \rightarrow P_2$ be a linear transformation for which

$$L(x+1) = x$$
 , $L(x-1) = 1$, $L(x^2) = 0$

Find
$$L(-x^2+3x+2)$$

First, write $\left(-x^2+3\ x+2\right)$ as a linear combination of (x+1) , (x-1), and $\left(x^2\right)$

$$\rightarrow (-x^2 + 3x + 2) = 2.5(x + 1) + 0.5(x - 1) - 1(x^2)$$

$$\rightarrow L(-x^2+3x+2)=2.5(x)+0.5(1)-1(0)$$

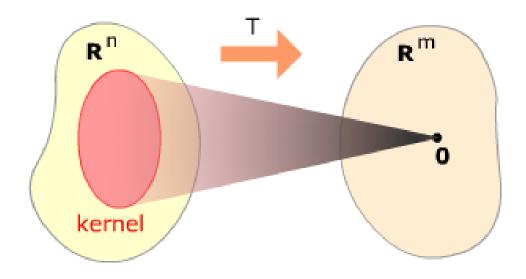
$$\rightarrow L(-x^2+3x+2)=2.5x+0.5$$

4. Kernal of the transformation (Null space)

Kernal of the transformation

It the set of all vectors that map into the zero vector upon transformation.

$$\ker(T) = \{x; T(x) = 0\}$$



Given $A_{m \times n}$ the matrix of transformation T

 $\ker(T)$ is the solution set of the homogeneous system $A \underset{n}{x} = 0 \rightarrow x \in \mathbb{R}^n$

 \rightarrow Also called "Nullspace" of A

Ex: Find the kernal of the transformation

$$T\left(\begin{bmatrix} x \\ y \\ z \end{bmatrix}\right) = \begin{bmatrix} x - y \\ x + z \end{bmatrix} \qquad \text{for } x = 1$$

 $\ker(T)$ is the set of all vectors x where T(x) = 0

$$\Rightarrow \begin{bmatrix} x - y \\ x + z \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix} \qquad \Rightarrow \begin{aligned} x - y &= 0 \\ \Rightarrow x + z &= 0 \end{aligned}$$

$$\rightarrow \begin{bmatrix} 1 & -1 & 0 \\ 1 & 0 & 1 \end{bmatrix} \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

AX=0 -... ang(A)

The matrix of transformation

$$\Rightarrow \begin{bmatrix} 1 & -1 & 0 & 0 \\ 1 & 0 & 1 & 0 \end{bmatrix} \sim \begin{bmatrix} 1 & -1 & 0 \\ 0 & 1 & 1 \end{bmatrix} \begin{bmatrix} 0 \\ 0 \end{bmatrix} \quad \text{Let } z = t \\ \text{E1: } x - (-t) = 0 \rightarrow x = -t \end{bmatrix} \text{ ker}(T) = t \begin{bmatrix} -1 \\ -1 \\ 1 \end{bmatrix}; t \in R$$

■ Thm : (Solutions of a homogeneous system)

If A is an $m \times n$ matrix, then the set of all solutions of the homogeneous system of linear equations $A\mathbf{x} = \mathbf{0}$ is a subspace of R^n called the nullspace of A.

Notes: The nullspace of A is also called the solution space of the homogeneous system $A\mathbf{x} = \mathbf{0}$.

Nullity:

The dimension of the nullspace of A is called the nullity of A.

$$\operatorname{nullity}(A) = \dim(NS(A))$$

Ex: (Finding the solution space of a homogeneous system)

Sol: The nullspace of A is the solution space of
$$Ax = 0$$
.

$$A = \begin{bmatrix} 1 & 2 & -2 & 1 \\ 3 & 6 & -5 & 4 \\ 1 & 2 & 0 & 3 \end{bmatrix} 0$$

$$0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

$$\Rightarrow x_1 = -2s - 3t, x_2 = s, x_3 = -t, x_4 = t$$

$$\Rightarrow x_{1} = -2s - 3t, x_{2} = s, x_{3} = -t, x_{4} = t$$

$$\Rightarrow \mathbf{x} = \begin{bmatrix} x_{1} \\ x_{2} \\ x_{3} \\ x_{4} \end{bmatrix} = \begin{bmatrix} -2s - 3t \\ s \\ -t \\ t \end{bmatrix} = s \begin{bmatrix} -2 \\ 1 \\ 0 \\ 0 \end{bmatrix} + t \begin{bmatrix} -3 \\ 0 \\ -1 \\ 1 \end{bmatrix} = s \mathbf{v}_{1} + t \mathbf{v}_{2}$$

$$\begin{cases} \mathbf{v}_{3} + \mathbf{v}_{2} + t \mathbf{v}_{3} = -t \\ \mathbf{v}_{3} + t = 0 \Rightarrow \mathbf{v}_{3} = -t \\ \mathbf{v}_{3} + t = 0 \Rightarrow \mathbf{v}_{3} = -t \\ \mathbf{v}_{1} + t = 0 \Rightarrow \mathbf{v}_{3} = -t \\ \mathbf{v}_{1} + t = 0 \Rightarrow \mathbf{v}_{2} = -t \\ \mathbf{v}_{1} + t = 0 \Rightarrow \mathbf{v}_{3} = -t \\ \mathbf{v}_{1} + t = 0 \Rightarrow \mathbf{v}_{3} = -t \\ \mathbf{v}_{1} + t = 0 \Rightarrow \mathbf{v}_{3} = -t \\ \mathbf{v}_{1} + t = 0 \Rightarrow \mathbf{v}_{3} = -t \\ \mathbf{v}_{1} + t = 0 \Rightarrow \mathbf{v}_{3} = -t \\ \mathbf{v}_{1} + t = 0 \Rightarrow \mathbf{v}_{3} = -t \\ \mathbf{v}_{1} + t = 0 \Rightarrow \mathbf{v}_{3} = -t \\ \mathbf{v}_{1} + t = 0 \Rightarrow \mathbf{v}_{3} = -t \\ \mathbf{v}_{1} + t = 0 \Rightarrow \mathbf{v}_{3} = -t \\ \mathbf{v}_{1} + t = 0 \Rightarrow \mathbf{v}_{3} = -t \\ \mathbf{v}_{1} + t = 0 \Rightarrow \mathbf{v}_{2} = -t \\ \mathbf{v}_{3} + t = 0 \Rightarrow \mathbf{v}_{3} = -t \\ \mathbf{v}_{1} + t = 0 \Rightarrow \mathbf{v}_{3} = -t \\ \mathbf{v}_{2} + t = 0 \Rightarrow \mathbf{v}_{3} = -t \\ \mathbf{v}_{3} + t = 0$$

$$\Rightarrow NS(A) = \{s\mathbf{v}_1 + t\mathbf{v}_2 \mid s, t \in R\}$$

$$\begin{bmatrix} 1 & 2 & -2 & 1 \end{bmatrix}$$

$$3 \ 6 \ -5 \ 4$$

$$X_1 = -25 - 3t$$

■ Thm : (Dimension of the solution space)

If *A* is an $m \times n$ matrix of rank *r*, then the dimension of the solution space of $A\mathbf{x} = \mathbf{0}$ is n - r. That is n = rank(A) + nullity(A)

• Notes:

(1) rank(A): The number of leading variables in the solution of Ax=0.

(The number of nonzero rows in the row-echelon form of A)

(2) nullity (A): The number of free variables in the solution of Ax = 0.

Ex: (Rank and nullity of a matrix)

Let the column vectors of the matrix A be denoted by \mathbf{a}_1 , \mathbf{a}_2 ,

$$A = \begin{bmatrix} 1 & 0 & -2 & 1 & 0 \\ 0 & -1 & -3 & 1 & 3 \\ -2 & -1 & 1 & -1 & 3 \\ 0 & 3 & 9 & 0 & -12 \end{bmatrix}$$

$$A = \begin{bmatrix} 1 & 0 & -2 & 1 & 0 \\ 0 & -1 & -3 & 1 & 3 \\ -2 & -1 & 1 & -1 & 3 \\ 0 & 3 & 9 & 0 & -12 \end{bmatrix}$$

$$A \times 5$$

- (a) Find the rank and nullity of *A*.
- (b) Find nullspace of A.

Sol: Let *B* be the reduced row-echelon form of *A*.

Fol: Let *B* be the reduced row-echelon form of *A*.

$$A = \begin{bmatrix} 1 & 0 & -2 & 1 & 0 \\ 0 & -1 & -3 & 1 & 3 \\ -2 & -1 & 1 & -1 & 3 \\ 0 & 3 & 9 & 0 & -12 \end{bmatrix}$$

$$B = \begin{bmatrix} 1 & 0 & -2 & 0 & 1 \\ 0 & 1 & 3 & 0 & -4 \\ 0 & 0 & 0 & 1 & -1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$
(a) rank(*A*) = 3 (the number of nonzero rows in *B*)

rank(A) = 3 (the number of nonzero rows in B)

nuillity(
$$A$$
) = n - rank(A) = $5 - 3 = 2$

(b) Find nullspace of A.

Let
$$x_3 = 5$$
, $x_5 = t$
E3: $x_4 - t = 0 \rightarrow x_4 = t$
E1: $x_2 + 35 - 4t = 0$
 $\rightarrow x_2 = -35 + 4t$
E1: $x_1 - 25 + t = 0 \rightarrow x_1 = 25 - t$

$$\begin{aligned}
xextT \\
y=NS(A) &= \begin{cases} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{cases} = \begin{cases} 2s-t \\ -3s+4t \\ t \end{cases} \\
&= \begin{cases} 2 \\ -3 \\ +t \end{cases} + \begin{cases} -1 \\ 4 \\ 0 \\ 1 \end{cases} + \begin{cases} 2s-t \\ -3s+4t \\ t \end{cases} \\
&= \begin{cases} 2s-t \\ -3s+4t \\ t \end{cases} \\
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