

# Math 244 - Linear Algebra

## Chapter 6: General Linear Transformations

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# Definition of a General Linear Transformation

## Definition

If  $T : V \longrightarrow W$  is a mapping from a vector space  $V$  to a vector space  $W$ , then  $T$  is called a **linear transformation** from  $V$  to  $W$  if the following two properties hold for all vectors  $\vec{u}$  and  $\vec{v}$  in  $V$  and for all scalars  $k$ :

①  $T(k\vec{u}) = kT(\vec{u})$  [Homogeneity property]

②  $T(\vec{u} + \vec{v}) = T(\vec{u}) + T(\vec{v})$  [Additivity property]

In the special case where  $V = W$ , the linear transformation  $T$  is called a **linear operator** on the vector space  $V$ .

## Examples of General Linear Transformations:

- Given a matrix  $A$  of size  $m \times n$ , the mapping  $T_A : \mathcal{M}_{n,1} \rightarrow \mathcal{M}_{m,1}$  defined by  $T_A(X) = AX$ , for  $X \in \mathcal{M}_{n,1}$  is a linear transformation.
- The mapping  $T : \mathcal{P}_3 \rightarrow \mathbb{R}^4$  defined by  $T(P) = (P(-1), P(0), P(1))$  is a linear transformation.
- The identity mapping  $Id : V \rightarrow V$  defined by  $Id(\vec{v}) = \vec{v}$ , for  $\vec{v} \in V$ , is a linear transformation.

# Properties of General Linear Transformations

## Remark

If  $T : V \longrightarrow W$  is a linear transformation, then:

$$T(k_1 \vec{u}_1 + k_2 \vec{u}_2) = k_1 T(\vec{u}_1) + k_2 T(\vec{u}_2)$$

for all  $\vec{u}_1$  and  $\vec{u}_2$  in  $V$  and  $k_1, k_2$  scalars.

More general:

$$T(k_1 \vec{u}_1 + k_2 \vec{u}_2 + \cdots + k_n \vec{u}_n) = k_1 T(\vec{u}_1) + k_2 T(\vec{u}_2) + \cdots + k_n T(\vec{u}_n)$$

for all  $\vec{u}_1, \vec{u}_2, \dots, \vec{u}_n$  in  $V$  and  $k_1, k_2, \dots, k_n$  scalars.

## Theorem

If  $T : V \longrightarrow W$  is a linear transformation, then:

- 1  $T(\vec{0}) = \vec{0}$ .
- 2  $T(\vec{u}_1 - \vec{u}_2) = T(\vec{u}_1) - T(\vec{u}_2)$  for all  $\vec{u}_1$  and  $\vec{u}_2$  in  $V$ .

# Kernel and Range of a Linear Transformation

## Definition

If  $T : V \longrightarrow W$  is a linear transformation, then the set of vectors in  $V$  that  $T$  maps into  $\vec{0}$  is called the kernel of  $T$  and is denoted by  $\ker(T)$ . The set of all vectors in  $W$  that are images under  $T$  of at least one vector in  $V$  is called the range of  $T$  and is denoted by  $R(T)$ .

## Theorem

*If  $T : V \longrightarrow W$  is a linear transformation, then:*

- 1 *The kernel of  $T$  is a subspace of  $V$ .*
- 2 *The range of  $T$  is a subspace of  $W$ .*

# Rank and nullity of a Linear Transformation

## Definition

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

## Theorem (Dimension Theorem for Linear Transformations)

*If  $T : V \longrightarrow W$  is a linear transformation from a finite-dimensional vector space  $V$  to a vector space  $W$ , then the range of  $T$  is finite-dimensional, and*

$$\text{rank}(T) + \text{nullity}(T) = \dim V.$$

# One-to-one Transformations

## Definition

If  $T : V \longrightarrow W$  is a linear transformation from a vector space  $V$  to a vector space  $W$ , then  $T$  is said to be **one-to-one** if  $T$  maps distinct vectors in  $V$  into distinct vectors in  $W$ .

## Theorem

*If  $T : V \longrightarrow W$  is a linear transformation, then the following statements are equivalent.*

- 1  $T$  is one-to-one.
- 2  $\ker(T) = \{\vec{0}\}$ .
- 3 If  $\{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_k\} \subset V$  is linearly independent, then  $\{T(\vec{v}_1), T(\vec{v}_2), \dots, T(\vec{v}_k)\} \subset W$  is linearly independent.

# Onto Transformations

## Definition

If  $T : V \longrightarrow W$  is a linear transformation from a vector space  $V$  to a vector space  $W$ , then  $T$  is said to be **onto** (or onto  $W$ ) if every vector in  $W$  is the image of at least one vector in  $V$ .

## Theorem

*If  $T : V \longrightarrow W$  is a linear transformation, then the following statements are equivalent.*

- 1  $T$  is onto.
- 2 If  $\{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_k\}$  is spanning  $V$ , then  $\{T(\vec{v}_1), T(\vec{v}_2), \dots, T(\vec{v}_k)\}$  is spanning  $W$ .



# One-to-one and Onto Transformations

## Theorem

*Let  $V$  and  $W$  be finite-dimensional vector spaces, and  $T : V \longrightarrow W$  a linear transformation.*

- ① *If  $T$  is one-to-one, then  $\dim V \leq \dim W$ .*
- ② *If  $T$  is onto, then  $\dim V \geq \dim W$ .*

## Theorem

*If  $V$  and  $W$  are finite-dimensional vector spaces with the same dimension, and if  $T : V \longrightarrow W$  is a linear transformation, then the following statements are equivalent.*

- ①  *$T$  is one-to-one.*
- ②  *$T$  is onto.*

## Definition

If  $T_1 : U \longrightarrow V$  and  $T_2 : V \longrightarrow W$  are linear transformations, then the composition of  $T_2$  with  $T_1$ , denoted by  $T_2 \circ T_1$  (which is read " $T_2$  circle  $T_1$ "), is the function defined by the formula

$$T_2 \circ T_1(\vec{u}) = T_2(T_1(\vec{u}))$$

where  $\vec{u}$  is a vector in  $U$ .

## Theorem

*If  $T_1 : U \longrightarrow V$  and  $T_2 : V \longrightarrow W$  are linear transformations, then  $T_2 \circ T_1 : U \longrightarrow W$  is also a linear transformation.*

Remark: If  $A$  and  $B$  are two matrices of size  $m \times n$  and  $n \times p$ , respectively, then  $T_A \circ T_B = T_{AB}$ .

# Inverse transformation and its properties

## Definition

An isomorphism is a one-to-one and onto linear transformation.

## Definition

Let  $T : U \longrightarrow V$  be a linear transformation. We say that  $T$  is invertible if there exists a mapping  $S : V \longrightarrow U$ , called an inverse of  $T$ , such that  $S \circ T = Id_U$  and  $T \circ S = Id_V$ .

## Theorem

*A linear transformation is invertible if and only if it is an isomorphism. In this case, the inverse of  $T$  is unique, denoted  $T^{-1}$  and is an isomorphism.*

# Inverse transformation and its properties

## Theorem

*Let  $T : U \longrightarrow V$  be an isomorphism of finite dimensional spaces  $U$  and  $V$ . We have:*

- ①  $\dim V = \dim W$  and  $(T^{-1})^{-1} = T$ .
- ② If  $T_1, T_2$  are invertible, then  $T_2 \circ T_1$  is invertible and we have  $(T_2 \circ T_1)^{-1} = T_1^{-1} \circ T_2^{-1}$ .

# Matrices for General Linear Transformations

## Theorem

Let  $T : V \rightarrow W$  be a linear transformation,  $B$  a basis of  $V$  and  $C$  a basis of  $W$ . There exists a matrix denoted  $[T]_{C,B}$  satisfying for any  $\vec{v} \in V$

$$[T(\vec{v})]_C = [T]_{C,B}[\vec{v}]_B.$$

If  $V = W$  and  $B = C$ , we write  $[T]_B$ .

Example: For  $T : \mathcal{P}_2 \rightarrow \mathcal{P}_2$  defined by

$T(P(X)) = P'(X) + P(0)X^2$  and  $B = C = (1, X, X^2)$ , we have for  $P(X) = aX^2 + bX + c$ ,

$$\begin{aligned} [T(P(X))]_C &= [P'(X) + P(0)X^2]_C = [2aX + b + cX^2]_C = \begin{pmatrix} b \\ 2a \\ c \end{pmatrix} \\ &= \begin{pmatrix} 0 & 1 & 0 \\ 0 & 0 & 2 \\ 1 & 0 & 0 \end{pmatrix} \begin{pmatrix} c \\ b \\ a \end{pmatrix} = [T]_{C,B}[P(X)]_B. \end{aligned}$$

# Matrices for General Linear Transformations

## Theorem

Let  $T : V \longrightarrow W$  be a linear transformation,  $B = \{\vec{v}_1, \vec{v}_2, \dots, \vec{v}_n\}$  a basis of  $V$  and  $C$  a basis of  $W$ . The matrix  $[T]_{C,B}$  associated to the linear transformation  $T$  with respect to the bases  $B$  and  $C$  is given by:

$$[T]_{C,B} = [[T(\vec{v}_1)]_C | T([\vec{v}_2)]_C | \cdots | [T(\vec{v}_n)]_C].$$

**Examples:** For  $T : \mathcal{P}_2 \longrightarrow \mathcal{P}_2$  defined by  $T(P(X)) = P'(X)$ , for all  $P(X) \in \mathcal{P}_2$ , and  $B = C = (1, X, X^2)$ , we have  $(1)' = 0$ ,

$$X' = 1, (X^2)' = 2X \text{ and } [T]_B = \begin{pmatrix} 0 & 1 & 0 \\ 0 & 0 & 2 \\ 0 & 0 & 0 \end{pmatrix}.$$

## Theorem

If  $T_1 : U \longrightarrow V$  and  $T_2 : V \longrightarrow W$  are linear transformations, and if  $B_1$ ,  $B_2$ , and  $B_3$  are bases for  $U$ ,  $V$ , and  $W$ , respectively, then  $[T_2 \circ T_1]_{B_3, B_1} = [T_2]_{B_3, B_2} [T_1]_{B_2, B_1}$ .

## Theorem

Let  $T : V \longrightarrow W$  be a linear transformation,  $B$  a basis of  $V$  and  $C$  a basis of  $W$ .

- 1  $T$  is one-to-one if and only if the matrix  $[T]_{C, B}$  has a left inverse.
- 2  $T$  is onto if and only if the matrix  $[T]_{C, B}$  has a right inverse.
- 3  $T$  is an isomorphism if and only if the matrix  $[T]_{C, B}$  is invertible. In this case  $[T^{-1}]_{B, C} = [T]_{C, B}^{-1}$ .

**Example:** Let  $T : \mathcal{P}_2 \longrightarrow \mathcal{P}_2$  be defined by

$T(P(X)) = XP'(X) + P(1)$  for  $P(X) \in \mathcal{P}_2$ .  $T$  is an isomorphism and we have  $T^{-1}(P(X)) = \int_1^X \frac{P(t)-P(0)}{t} dt + P(0)$  for  $P(X) \in \mathcal{P}_2$ . Consider  $B = (1, X, X^2)$ .

Because  $T(1) = 1$ ,  $T(X) = X + 1$  and  $T(X^2) = 2X^2 + 1$ , we find

$$[T]_B = \begin{pmatrix} 1 & 1 & 1 \\ 0 & 1 & 0 \\ 0 & 0 & 2 \end{pmatrix}.$$

Because  $T^{-1}(1) = 1$ ,  $T^{-1}(X) = X - 1$  and  $T^{-1}(X^2) = \frac{1}{2}X^2 - \frac{1}{2}$ ,

we find  $[T^{-1}]_B = \begin{pmatrix} 1 & -1 & -\frac{1}{2} \\ 0 & 1 & 0 \\ 0 & 0 & \frac{1}{2} \end{pmatrix}.$

Clearly  $[T]_B[T^{-1}]_B = I$ . Thus  $[T^{-1}]_B = [T]_B^{-1}$ .



## Theorem

Let  $T : V \rightarrow W$  be a linear transformation,  $B$  and  $B'$  two bases of  $V$  and  $C$  and  $C'$  two bases of  $W$ . We have

$$[T]_{C',B'} = P_{C' \leftarrow C} [T]_{C,B} P_{B \leftarrow B'}.$$

**Example:** Let  $T : \mathcal{P}_2 \rightarrow \mathcal{P}_2$  be defined for  $P(X) \in \mathcal{P}_2$  by  $T(P(X)) = XP'(X) + P(1)$ . Let  $B = (X^2 - 1, X^2 - X, X^2 + X)$  and  $B' = (1, X, X^2)$ .

Because  $T(X^2 - 1) = 2X^2 = (X^2 - X) + (X^2 + X)$ ,  
 $T(X^2 - X) = 2X^2 - X = \frac{3}{2}(X^2 - X) + \frac{1}{2}(X^2 + X)$  and  
 $T(X^2 + X) = 2X^2 + X + 2 = -2(X^2 - 1) + \frac{3}{2}(X^2 - X) + \frac{5}{2}(X^2 + X)$ ,

we find  $[T]_B = \begin{pmatrix} 0 & 0 & -2 \\ 1 & \frac{3}{2} & \frac{3}{2} \\ 1 & \frac{1}{2} & \frac{5}{2} \end{pmatrix}$ . Moreover  $[T]_{B'} = \begin{pmatrix} 1 & 1 & 1 \\ 0 & 1 & 0 \\ 0 & 0 & 2 \end{pmatrix}$ ,

$$P_{B' \leftarrow B} = \begin{pmatrix} -1 & 0 & 0 \\ 0 & -1 & 1 \\ 1 & 1 & 1 \end{pmatrix}, \quad P_{B \leftarrow B'} = P_{B' \leftarrow B}^{-1} = \frac{1}{2} \begin{pmatrix} -2 & 0 & 0 \\ 1 & -1 & 1 \\ 1 & 1 & 1 \end{pmatrix}.$$

We check and find that  $[T]_{B'} = P_{B' \leftarrow B} [T]_B P_{B \leftarrow B'}$ .