

# Week 5 Linear Transformations and Matrices

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# 1 Linear Transformations

## Definition 1.1: Linear Transformation

A function  $T : V \rightarrow W$  is called linear if

$$\#1 \quad T(x + y) = T(x) + T(y)$$

$$\#2 \quad T(cx) = cT(x)$$

## Proposition 1.1

A map  $T : V \rightarrow W$  is linear **if and only if**

$$T(cx + y) = cT(x) + T(y)$$

**Remark 1.1.** The transformation  $T : M_{m \times n}(\mathbb{F}) \rightarrow M_{n \times m}(\mathbb{F})$  defined by  $T(A) = A^t$  (Transpose) is a linear transformation. The transpose is defined as

$$A_{ij}^t = A_{ji}$$

**Remark 1.2.** if  $T$  is linear then

- $T(0) = 0$

## Theorem 1.2: Unique Linear Transformation

Let  $\{v_1, \dots, v_n\}$  be a basis for a vector space  $V$  then let  $\{w_1, \dots, w_n\} \in W$  then there exists a **unique** linear transformation such that

$$T(v_1) = w_1 \cdots T(v_n) = w_n$$

## Definition 1.2: Null Space and Range

The null space or kernel of a linear transformation is defined as

$$N(T) = \{v \in V \mid T(v) = 0_W\}$$

The range is defined as

$$R(T) = \{T(v) \mid v \in V\}$$

**Remark 1.3.**  $N(T)$  is a subspace of  $V$  and  $R(T)$  is a subspace of  $W$ .

### Theorem 1.3: Rank-Nullity Theorem

$$\text{Rank}(T) + \text{Null}(T) = \dim V$$

*proof sketch.* Extending a basis  $\{v_i\}_{i=1}^k$  of  $N(T)$  to form a basis of  $V$  and removing forming a basis for  $R(T)$  with  $n - k$  vectors defined by  $\{T(v_j)\}_{j=k+1}^n$ .  $\square$

### Definition 1.3: Isomorphism

A linear transformation is also called an isomorphism if it is also a bijection.

### Lemma 1.4

Let  $T$  be linear then  $T$  is injective if and only if  $N(T) = \{0_V\}$

### Theorem 1.5: Basis for Isomorphic vector Spaces

Let  $T : V \rightarrow W$  be linear. Let  $\{v_1, \dots, v_n\}$  be a basis for  $V$  then  $T$  is an isomorphism if and only if  $\{T(v_1), \dots, T(v_n)\}$  is a basis for  $W$

### Theorem 1.6: Isomorphism Condition

$V \cong W$  if and only if  $\dim V = \dim W$  for  $\dim V, W < \infty$

### Theorem 1.7: Equivalent Conditions

Let  $\dim W, \dim V < \infty$  and  $\dim V = \dim W$ . If  $T : V \rightarrow W$  is linear then the following are **equivalent**

- (1)  $T$  is one-to-one.
- (2)  $T$  is onto.
- (3)  $\text{rank}(T) = \dim V$ .

## 2 Matrices and Linear Transformations

### 2.0.1 Set of all linear transformations

#### Definition 2.1: $\mathcal{L}$

Let  $V, W$  be vector spaces then  $\mathcal{L}(V, W)$  is the set of all linear transformation  $T : V \rightarrow W$ .

**Remark 2.1.** For any sets  $A, B$  the set  $A^B$  is the set of all functions  $B \rightarrow A$ . We can define the operations on this set by

$$(f + g)(x) = f(x) + g(x)$$

$$(cf)(x) = cf(x)$$

If  $A, B$  are vector spaces then  $A^B$  is also a vector space with these operations.

#### Theorem 2.1

$\mathcal{L}(V, W)$  is a subspace of  $W^V$

#### Definition 2.2: Matrix-Vector Multiplication

For a column vector  $\mathbf{x} = (x_1, x_2, \dots, x_n)^T$  and the matrix

$$A = \begin{bmatrix} a_{11} & a_{12} & \dots \\ a_{21} & a_{22} & \dots \\ \vdots & \ddots & \vdots \\ a_{m1} & \dots & a_{mn} \end{bmatrix}$$

The product  $A\mathbf{x}$  is defined as

$$\begin{bmatrix} \sum_{i=1}^n a_{1i}x_i \\ \sum_{i=1}^n a_{2i}x_i \\ \vdots \\ \sum_{i=1}^n a_{mi}x_i \end{bmatrix}$$

**Remark 2.2.** We can also define the product  $A\mathbf{x}$  in terms of linear combination of the column vectors of  $A$

$$x_1 \begin{bmatrix} a_1 \\ \vdots \\ a_m \end{bmatrix} + x_2 \begin{bmatrix} a_2 \\ \vdots \\ a_m \end{bmatrix} + \dots + x_n \begin{bmatrix} a_n \\ \vdots \\ a_m \end{bmatrix}$$

Where  $a_i$  is the  $i$ -th column vector. The result is a  $m \times 1$  vector. For  $e_j \in \mathbb{F}^n$  the product  $Ae_j$  is the  $j$ -th column vector of  $A$ .

**Definition 2.3:  $L_A$  function**

Let  $A \in M_{m \times n}(\mathbb{F})$  then  $L_A$  denotes the function  $\mathbb{F}^n \rightarrow \mathbb{F}^m$  defined by

$$L_A(\mathbf{x}) = A\mathbf{x}$$

**Corollary 2.0.1: Matrix Equality Theorem**

Let  $A, B \in M_{m \times n}(\mathbb{F})$  then  $A = B$  if and only if for all  $\mathbf{x} \in \mathbb{F}^n$  we have

$$A\mathbf{x} = B\mathbf{x}$$

**Theorem 2.2: Linearity of Matrix Vector Multiplication**

Let  $A \in M_{m \times n}(\mathbb{F})$  then the linear transformation  $L_A : \mathbb{F}^n \rightarrow \mathbb{F}^m$  is linear

*proof sketch.* Prove using  $L_A(c\mathbf{x} + \mathbf{y}) = cL_A(\mathbf{x}) + L_A(\mathbf{y})$  and the linear combination of column vector version of matrix vector multiplication. □

**Proposition 2.3**

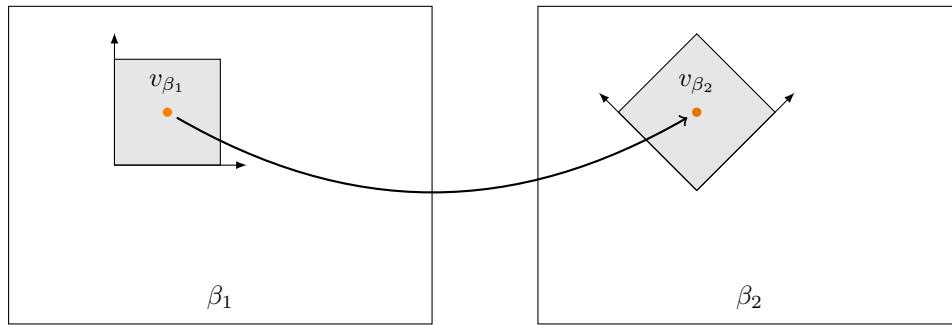
Let  $\mathbf{L} : M_{m \times n}(\mathbb{F}) \rightarrow \mathcal{L}(\mathbb{F}^n, \mathbb{F}^m)$  be defined as

$$\mathbf{L}(A) = L_A$$

Then  $\mathbf{L}$  is linear and one-to-one. (Injective).

**Remark 2.3.**  $\mathbf{L}$  is actually an isomorphism from  $M_{m \times n}(\mathbb{F}) \rightarrow \mathcal{L}(\mathbb{F}^n, \mathbb{F}^m)$

### 3 Coordinates



#### Definition 3.1: Ordered Basis

An **ordered basis** for  $V$  is a basis set  $\{v_1, \dots, v_n\}$  with a specific order.

#### Definition 3.2: Coordinate Vector

Let  $\beta$  be an **ordered basis** for  $V$  then the **coordinate vector** for  $x \in V$  is  $[x]_\beta$  is

$$[x]_\beta = \begin{bmatrix} a_1 \\ a_2 \\ \vdots \\ a_n \end{bmatrix} \in \mathbb{F}^n$$

Where  $x = \sum_{k=1}^n a_k v_k$  where  $v_k \in \beta$ .

#### Theorem 3.1

The map  $[\ ]_\beta : V \rightarrow \mathbb{F}^n$  where  $\dim V = n$  is an isomorphism.

## 4 Matrix Representation of a Linear Transformation

### Definition 4.1

Let  $V, W$  be vector spaces over  $\mathbb{F}$ . Let  $T : V \rightarrow W$  be linear and  $\beta = \{v_1, \dots, v_n\}$  be an ordered basis for  $V$  and  $\gamma$  be an ordered basis for  $W$ . Then the matrix representation of  $T$  is given by

$$[T]_{\beta}^{\gamma} = \begin{bmatrix} [T(v_1)]_{\gamma} & [T(v_2)]_{\gamma} & \cdots & [T(v_n)]_{\gamma} \end{bmatrix}$$

If  $T : V \rightarrow V$  then  $[T]_{\beta}$  denotes  $[T]_{\beta}^{\beta}$ .

**Remark 4.1.** If  $A = [T]_{\beta}^{\gamma}$  then  $A \in M_{m \times n}(\mathbb{F})$  where  $\dim W = m$  and  $\dim V = n$ .

### Theorem 4.1

Let  $T : V \rightarrow W$  be linear and  $\beta, \gamma$  be ordered basis for  $V, W$  then

$$[T(x)]_{\gamma} = [T]_{\beta}^{\gamma} [x]_{\beta}$$

### Proposition 4.2

(1) For  $T, U \in \mathcal{L}(V, W)$  we have

$$[cT + U]_{\beta}^{\gamma} = c[T]_{\beta}^{\gamma} + [U]_{\beta}^{\gamma}$$

(2) For every  $C \in M_{m \times n}(\mathbb{F})$  there exists  $T \in \mathcal{L}(V, W)$  such that  $[T]_{\beta}^{\gamma} = C$ .

**Remark 4.2.**  $\left[ \begin{array}{c} \\ \end{array} \right]_{\beta}^{\gamma}$  is an isomorphism  $\mathcal{L}(V, W) \rightarrow M_{m \times n}(\mathbb{F})$ . Where  $m = \dim W$  and  $n = \dim V$ .

## 5 Matrix Multiplication

### Definition 5.1: Matrix Multiplication

Let  $A \in M_{m \times n}$  and  $B \in M_{n \times p}$  then the product  $AB = C \in M_{m \times p}$  is defined by

$$C_{ij} = \sum_{t=1}^n a_{it}b_{tj}$$

If numbers of columns of  $A$  does not equal number of rows in  $B$  then  $AB$  is not defined.

**Remark 5.1.** We can also compute  $AB$  by considering the products  $Ab_i$  where  $b_i$  is the  $i$ -th column vector of  $B$ , then

$$AB = \begin{bmatrix} Ab_1 & Ab_2 & \cdots & Ab_p \end{bmatrix}$$

### Lemma 5.1: Properties of matrix Multiplication

1.  $A(B + C) = AB + AC$  where  $A \in M_{m \times n}$  and  $B, C \in M_{n \times p}$
2.  $(D + E)A = DA + EA$  where  $D, E \in M_{m \times n}$  and  $A \in M_{n \times p}$
3.  $\alpha(AB) = (\alpha A)B = A(\alpha B)$  where  $\alpha \in \mathbb{F}$
4.  $(AB)^t = B^t A^t$
5.  $I_m A = A I_n = A$
6.  $A O_{n \times p} = O_{m \times p}$  and  $O_{q \times m} A = O_{q \times n}$

### Theorem 5.2: Matrix of Composition of Linear Transformations

Let  $V, W, Z$  be finite dimensional vector spaces. With ordered bases  $\alpha = \{v_1, \dots, v_p\}$ ,  $\beta = \{w_1, \dots, w_n\}$  and  $\gamma = \{z_1, \dots, z_m\}$ . Now let

$$T : V \rightarrow W \quad U : W \rightarrow Z$$

Be linear. Then the corresponding matrices are  $[T]_{\alpha}^{\beta} \in M_{m \times p}$  and  $[U]_{\beta}^{\gamma} \in M_{n \times p}$  and let  $C = [UT]_{\alpha}^{\gamma} \in M_{m \times p}$ . Then  $C = AB$ , that is

$$[UT]_{\alpha}^{\gamma} = [U]_{\beta}^{\gamma} [T]_{\alpha}^{\beta}$$

*Proof sketch.*  $[U]_{\beta}^{\gamma} [T(v_j)]_{\beta}$  is the  $j$ -th column of  $[U]_{\beta}^{\gamma} [T]_{\alpha}^{\beta}$  and  $[U]_{\beta}^{\gamma} [T(v_j)]_{\beta} = [U(x)]_{\gamma}$  where  $x = T(v_j)$ . Therefore  $[U(T(v_j))]_{\gamma} = [UT(v_j)]_{\gamma}$ . So both matrices have the same  $j$ -th column.  $\square$



**Proposition 5.3**

Let  $A \in M_{m \times n}(\mathbb{F})$  and consider  $L_A : \mathbb{F}^n \rightarrow \mathbb{F}^m$ . Let  $\beta$  be the [standard ordered basis](#) for  $\mathbb{F}^n$  and let  $\gamma$  be the standard ordered basis for  $\mathbb{F}^m$  then

$$[L_A]_{\beta}^{\gamma} = A$$

**Corollary 5.0.1**

#1  $L_{AB} = L_A L_B$  where  $A \in M_{m \times n}$  and  $B \in M_{n \times p}$

#2  $A(BC) = (AB)C$

*Proof sketch.* Let  $\alpha$  be standard ordered basis for  $\mathbb{F}^p$ ,  $\beta$  be the standard ordered basis for  $\mathbb{F}^n$  and let  $\gamma$  be the standard ordered basis for  $\mathbb{F}^m$ . Then we know that  $[L_A B]_{\alpha}^{\gamma} = AB$  and

$$\begin{aligned} [L_A L_B]_{\alpha}^{\gamma} &= [L_A]_{\alpha}^{\beta} [L_B]_{\alpha}^{\beta} \\ &= AB \end{aligned}$$

Since  $[ ]_{\alpha}^{\beta}$  is one-to-one we get  $L_A L_B = L_{AB}$ .

(#2) Using the function  $L$  defined [Proposition 2.3](#) since it's injective,  $L_{ABC} = L_A L_B L_C$  and function composition is associative. □