

# Linear Algebra

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# Lecture 26

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# Linear Transformation

## Proposition 28

Two finite-dimensional vector spaces  $V$  and  $W$  (over the same field  $\mathbb{F}$ ) are isomorphic if and only if  $\dim V = \dim W$ .

**Remark:** In particular, it follows that every vector space  $V$  of dimension  $n$  over  $\mathbb{R}$  is isomorphic to  $\mathbb{R}^n$ . We can certainly exploit our familiarity with  $\mathbb{R}^n$  in proving results about finite-dimensional spaces.

### Proof of Proposition 28:

[ $\implies$ ]

Suppose  $V$  and  $W$  are isomorphic, i.e., there exists an isomorphism  $T : V \longrightarrow W$ . If  $\dim V = n$  and  $\{v_1, \dots, v_n\}$  is a basis of  $V$ . Then by Proposition 27 (a),  $\{T(v_1), \dots, T(v_n)\}$  is a basis of  $W$ , and hence,  $\dim W = n$ .

[ $\impliedby$ ]

Suppose  $V$  and  $W$  have the same dimension, say  $n$ . Let  $\{v_1, \dots, v_n\}$  be a basis of  $V$  and let  $\{w_1, \dots, w_n\}$  be a fixed basis of  $W$ . By Prop 26(b), there exists a unique linear transformation  $T : V \longrightarrow W$  such that  $T(v_i) = w_i, i = 1, \dots, n$ . But now, from Prop 27(b), it follows that  $T$  is an isomorphism.

# Rank of a Linear Transformation

**Definition:** If  $T : V \longrightarrow W$  be a linear transformation, then  $\text{Range}(T) = \{w \in W : w = T(v) \text{ for some } v \in V\}$ —this is the standard definition of the range of any function.

**Remark:**  $\text{Range}(T)$  is a subspace of  $W$  (exercise!)

**Definition:** The **rank** of  $T$  is the dimension of the range of  $T$ .

**Remark:** If  $V$  is finite-dimensional, then it is easy to see that  $\text{rank}(T) \leq \dim V$  (exercise!!)

**Recall:** We have already defined the kernel or null space of a linear transformation  $T$  as  $\text{Ker}(T) = \text{Nul}(T) = \{v \in V : T(v) = 0\}$  and see that it is a subspace of  $V$ . If  $\text{Ker } T$  is finite-dimensional, then its dimension is called the **nullity** of  $T$ .

**Theorem 3 (Rank-Nullity Theorem for Linear Transformation):**

Suppose that  $T : V \longrightarrow W$  is a linear transformation and  $V$  is finite-dimensional. Then:

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

## Proof of Theorem 3 (Rank-Nullity Theorem)

Suppose that  $\dim V = n$  and  $\text{nullity}(T) = k$ . Then let  $\{v_1, v_2, \dots, v_k\}$  be a basis for  $\text{Nul } T$  and expand this to a basis  $B$  of  $V$  by inserting the additional vectors  $v_{k+1}, \dots, v_n$ .

**Claim:** The vectors  $T(v_{k+1}), \dots, T(v_n)$  form a basis for  $\text{Range}(T)$ .

Firstly, all the vectors  $T(v_1), \dots, T(v_n)$  surely span  $\text{Range}(T)$ , and since  $T(v_1) = T(v_2) = \dots = T(v_k) = 0$ , actually  $T(v_{k+1}), \dots, T(v_n)$  span  $\text{Range}(T)$ .

Secondly, suppose that  $c_{k+1}T(v_{k+1}) + c_{k+2}T(v_{k+2}) + \dots + c_nT(v_n) = 0$

Then  $T(c_{k+1}v_{k+1} + c_{k+2}v_{k+2} + \dots + c_nv_n) = 0$ . Hence,

$c_{k+1}v_{k+1} + c_{k+2}v_{k+2} + \dots + c_nv_n$  belongs to  $\text{Nul } T$ . Therefore

$c_{k+1}v_{k+1} + c_{k+2}v_{k+2} + \dots + c_nv_n = b_1v_1 + b_2v_2 + \dots + b_kv_k$  or

$b_1v_1 + b_2v_2 + \dots + b_kv_k - c_{k+1}v_{k+1} - c_{k+2}v_{k+2} - \dots - c_nv_n = 0$ , which implies that  $b_1 = b_2 = \dots = b_k = c_{k+1} = c_{k+2} = \dots = c_n = 0$ . In

particular,  $c_{k+1} = c_{k+2} = \dots = c_n = 0$ , which proves the linear independence of the vectors  $T(v_{k+1}), \dots, T(v_n)$ . This proves the claim.

Hence  $\text{Rank } T + \text{Nullity } T = (n - k) + k = n = \dim V$ .

**Observation:** We have already noted that if  $T$  is a linear transformation on a finite-dimensional space  $V$ , then  $\text{range } T$  is also finite-dimensional of dimension at most  $\dim V$ . The Rank-Nullity Theorem makes this numerically precise.

# Review – Basis and Dimension

- **Definition:** Let  $V$  be a vector space. A basis for  $V$  is a linearly independent set  $S$  of vectors such that  $V = \text{Span } S$ .
- **Definition:** A space  $V$  which has a (finite) basis is said to be finite dimensional.
- **Proposition 13:** If  $V$  is a finite-dimensional vector space, then any two bases of  $V$  have the same number of elements.
- **Definition:** The dimension of a finite-dimensional space is the number of elements in a basis for  $V$ . This is written  $\dim V$ .
- Three examples of finite dimensional spaces:  $\mathbb{R}^n, \mathbb{R}^{m \times n}, \mathbb{R}_n[t]$ . Their dimensions are  $n, mn, n + 1$ , respectively.



# Very Important Theorem (VIT)–Version 2.0

## The Invertible Matrix Theorem (TIMT)

**Theorem 1:** The following are equivalent for an  $m \times m$  square matrix  $A$ :

- (a)  $A$  is invertible.
- (b)  $A$  is row equivalent to the identity matrix  $I_m$ .
- (c) The homogeneous system  $AX = 0$  has only the trivial solution.
- (d) The system of equations  $AX = b$  has at least one solution for every  $b$  in  $\mathbb{R}^m$ .
- (e) Nullity  $(A) = 0$ .
- (f) Rank  $(A) = m$ .
- (g) The columns of  $A$  form a basis for  $\mathbb{R}^m$ .
- (h)  $\text{Det } A \neq 0$ .