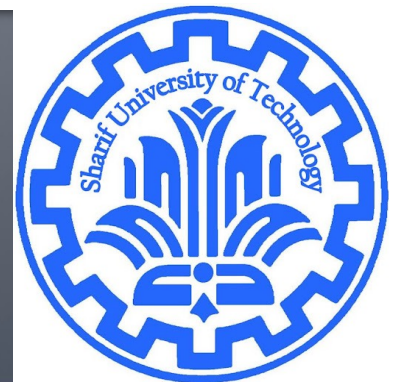


# Matrix Algebra: Dimension and Rank

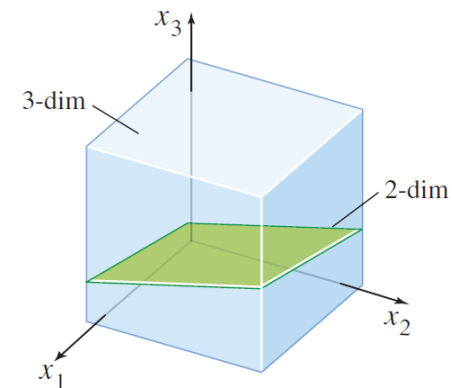
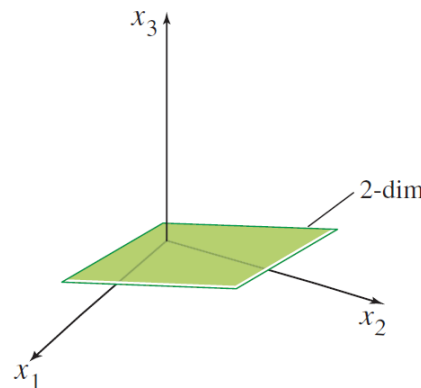
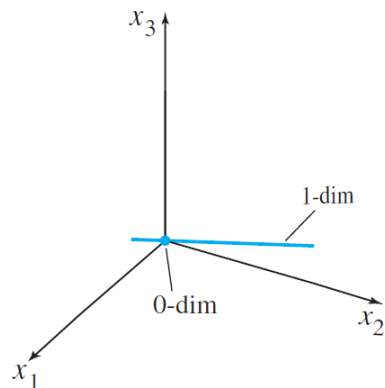
CE40282-1: Linear Algebra  
Hamid R. Rabiee and Maryam Ramezani  
Sharif University of Technology



# Review: Dimension

- If  $V$  has a finite basis, then  $\dim(V)$  is the number of elements (vectors) of any basis of  $V$
- $\dim(\{0\})=0$

If  $V$  is spanned by a finite set, then  $V$  is said to be **finite-dimensional**, and the **dimension** of  $V$ , written as  $\dim V$ , is the number of vectors in a basis for  $V$ . The dimension of the zero vector space  $\{0\}$  is defined to be zero. If  $V$  is not spanned by a finite set, then  $V$  is said to be **infinite-dimensional**.



# Finite-Dimensional Space

## THEOREM

Let  $H$  be a subspace of a finite-dimensional vector space  $V$ . Any linearly independent set in  $H$  can be expanded, if necessary, to a basis for  $H$ . Also,  $H$  is finite-dimensional and

$$\dim H \leq \dim V$$

## THEOREM

### The Basis Theorem

Let  $V$  be a  $p$ -dimensional vector space,  $p \geq 1$ . Any linearly independent set of exactly  $p$  elements in  $V$  is automatically a basis for  $V$ . Any set of exactly  $p$  elements that spans  $V$  is automatically a basis for  $V$ .

# Row and Column Space

## THEOREM

If two matrices  $A$  and  $B$  are row equivalent, then their row spaces are the same. If  $B$  is in echelon form, the nonzero rows of  $B$  form a basis for the row space of  $A$  as well as for that of  $B$ .

The pivot columns of a matrix  $A$  form a basis for  $\text{Col } A$ .

# Row and Column Space

- Example

- Row Basis
- Column Basis
- $\dim(\text{row}(A))$
- $\dim(\text{column}(A))$
- $\dim(\text{null}(A))$

$$A = \begin{bmatrix} -2 & -5 & 8 & 0 & -17 \\ 1 & 3 & -5 & 1 & 5 \\ 3 & 11 & -19 & 7 & 1 \\ 1 & 7 & -13 & 5 & -3 \end{bmatrix}$$

$$A \sim B = \begin{bmatrix} 1 & 3 & -5 & 1 & 5 \\ 0 & 1 & -2 & 2 & -7 \\ 0 & 0 & 0 & -4 & 20 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

$$A \sim B \sim C = \begin{bmatrix} 1 & 0 & 1 & 0 & 1 \\ 0 & 1 & -2 & 0 & 3 \\ 0 & 0 & 0 & 1 & -5 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

number of non-zero rows=pivot columns

# Rank

- Rank of matrix: (row or column)
    - The number of linearly independent rows or columns in the matrix
    - Dimension of the row (column) space
    - Number of nonzero rows of the matrix in row echelon form (Ref)
- 1) Row rank = column rank for a matrix in reduced row echelon form.
- 2) The dimension of the column space of  $A$  and  $\text{rref}(A)$  is the same.

# Rank

**Theorem RMRT: Rank of a Matrix is the Rank of the Transpose.** Suppose  $A$  is an  $m \times n$  matrix. Then  $r(A) = r(A^t)$

# Range Space

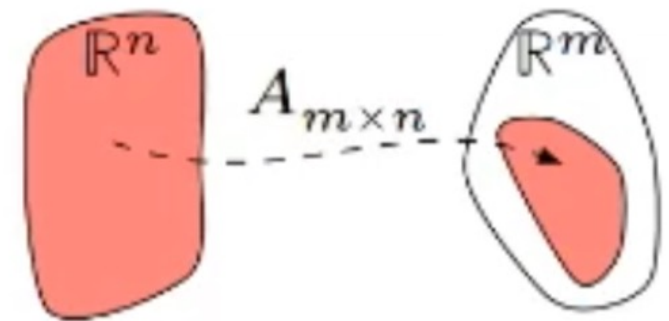
- For  $A_{m \times n} = [a^1 \dots a^n] = [a_1 \dots a_n]$

$$\text{range}(A) = \text{span}(a_1, \dots, a_n)$$

$$= \left\{ y \mid y = \alpha_1 a_1 + \dots + \alpha_n a_n, \alpha_1, \dots, \alpha_n \in \mathbb{R} \right\}$$

$$= \left\{ y \mid y = Ax, x \in \mathbb{R}^n \right\}$$

- Range is a vector space
- Range of  $A$  is a subspace of  $\mathbb{R}^m$
- Is  $\text{Dim}(A)=m$ ?



$$\dim(\text{range}(A)) = \text{colrank}(A)$$

number of linear independent columns

- Example

$$A = \begin{bmatrix} 0 & 1 & 1 & 2 \\ 1 & 0 & 2 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$



# Null space (kernel)

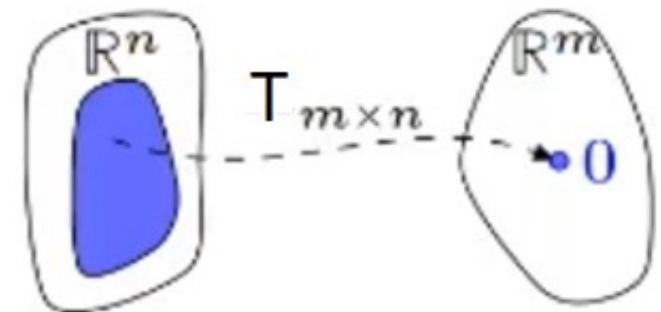
- Definition

Let  $T : V \rightarrow W$  be a linear map. Then the **null space** or **kernel** of  $T$  is the set of all vectors in  $V$  that map to zero:

$$N(T) = \text{null } T = \{v \in V \mid Tv = 0\}.$$

- Null space is a vector space
- Null space  $T$  is a subspace of  $(V) \mathbb{R}^n$
- Is  $\text{Dim}(\text{null } (T))=n$ ?
- Nullity( $T$ ):  $\text{Dim}(\text{null } (T))$
- Question:

- What is null space for differentiation mapping?



# Null space (kernel)

- Nullity(A)=the number of free variables
- Example
  - If columns of matrix (A) are linearly independent  
 $\text{nullity}(A) = ?$   
 $\text{col}(\text{rank}(A)) = ?$

$$\blacksquare A = \begin{bmatrix} 0 & 1 & 1 & 2 \\ 1 & 0 & 2 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

$$x = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix}, Ax = \begin{bmatrix} x_2 + x_3 + 2x_4 \\ x_1 + 2x_3 + x_4 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \Rightarrow x = \begin{bmatrix} -2x_3 - x_4 \\ -x_3 - 2x_4 \\ x_3 \\ x_4 \end{bmatrix} = x_3 \begin{bmatrix} -2 \\ -1 \\ 1 \\ 0 \end{bmatrix} + x_4 \begin{bmatrix} -1 \\ -2 \\ 0 \\ 1 \end{bmatrix}$$

$$\text{nullity}(A) = 2 \quad \text{col rank}(A) = 2$$

# Conclusion

**Theorem 2.7.17.** *The vectors attached to the free variables in the parametric vector form of the solution set of  $Ax = 0$  form a basis of  $\text{Nul}(A)$ .*

Let  $A$  be an  $n \times n$  matrix. Then the following statements are each equivalent to the statement that  $A$  is an invertible matrix.

- m. The columns of  $A$  form a basis of  $\mathbb{R}^n$ .
- n.  $\text{Col } A = \mathbb{R}^n$
- o.  $\dim \text{Col } A = n$
- p.  $\text{rank } A = n$
- q.  $\text{Nul } A = \{\mathbf{0}\}$
- r.  $\dim \text{Nul } A = 0$

# Conclusion

The dimension of  $\text{Nul } A$  is the number of free variables in the equation  $A\mathbf{x} = \mathbf{0}$ , and the dimension of  $\text{Col } A$  is the number of pivot columns in  $A$ .

- Examples:
  - Go to slide 6
  - Find the dimensions of the null space and the column space of

$$A = \begin{bmatrix} -3 & 6 & -1 & 1 & -7 \\ 1 & -2 & 2 & 3 & -1 \\ 2 & -4 & 5 & 8 & -4 \end{bmatrix}$$

Row reduce the augmented matrix  $[A \ \mathbf{0}]$  to echelon form

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

# Rank-Nullity Theorem

- $\text{nullity}(A) + \text{col rank}(A) = n$
- $\dim(\text{null}(A)) + \dim(\text{range}(A)) = n$ 
  - Proof?

$$\left\{ \begin{array}{c} \text{number of} \\ \text{pivot columns} \end{array} \right\} + \left\{ \begin{array}{c} \text{number of} \\ \text{nonpivot columns} \end{array} \right\} = \left\{ \begin{array}{c} \text{number of} \\ \text{columns} \end{array} \right\}$$

# Rank Theorem

- Theorem:
  - $\text{col rank}(A) = \text{row rank}(A)$
  - In general it is called rank of matrix!  $\text{rank}(A)$
  - Proof?

# Rank Properties

- $\text{col rank}(A_{m \times n}) \leq \min(m, n)$
- $\text{row rank}(A_{m \times n}) \leq \min(m, n)$
- $\dim(\text{range}(A)) = \text{rank}(A)$   
 $\text{nullity}(A) + \text{rank}(A) = n$   
 $\text{rank}(A) \leq \min(m, n)$

# Rank Properties

- For  $A, B \in \mathbb{R}^{m \times n}$ 
  1.  $\text{rank}(A) \leq \min(m, n)$
  2.  $\text{rank}(A) = \text{rank}(A^T)$
  3.  $\text{rank}(AB) \leq \min(\text{rank}(A), \text{rank}(B))$
  4.  $\text{rank}(A + B) \leq \text{rank}(A) + \text{rank}(B)$
- $A$  has *full rank* if  $\text{rank}(A) = \min(m, n)$
- If  $m > \text{rank}(A)$  rows not linearly independent
  - Same for columns if  $n > \text{rank}(A)$



# Rank Properties

- The **range** or the column space of a matrix  $A \in \mathbb{R}^{m \times n}$ , denoted  $\mathcal{R}(A)$ , is the span of the columns of  $A$ . In other words,

$$\mathcal{R}(A) = \{v \in \mathbb{R}^m : v = Ax, x \in \mathbb{R}^n\}.$$

- Assuming  $A$  is full rank and  $n < m$ , the projection of a vector  $y \in \mathbb{R}^m$  onto the range of  $A$  is given by,

$$\text{Proj}(y; A) = \operatorname{argmin}_{v \in \mathcal{R}(A)} \|v - y\|_2 = A(A^T A)^{-1} A^T y \ .$$

- When  $A$  contains only a single column,  $a \in \mathbb{R}^m$ , this gives the special case for a projection of a vector onto a line:

$$\text{Proj}(y; a) = \frac{aa^T}{a^T a} y \ .$$