

# 7\_5. The rank theorem

Column rank = row rank.

A deep thought indeed!

# The rank theorem

**Theorem 7.5.1** (*The Rank Theorem*) *The row space and column space of a matrix have the same dimension.*

- Proof: Let  $T: \mathbb{R}^m \rightarrow \mathbb{R}^n$  be defined by  $T(x) = Ax$ . Then
  - $\dim \text{ran } T = \dim \text{column space } A$  since  $v = x_1 A_1 + \dots + x_n A_n$  for  $v$  in  $\text{ran } T$  and  $A_i$  column vectors.
  - $\text{Ker } T = \text{null } A$ .
  - $\dim \text{ran } T + \dim \text{ker } T = n$ .
    - Choose a basis  $a_1, \dots, a_k$  in  $\text{ker } T$ .
    - Expand  $a_{k+1}, \dots, a_n$  in  $\mathbb{R}^n$  to a basis.
    - $T(a_{k+1}), \dots, T(a_n)$  is independent. They span  $\text{ran } T$ .
    - Thus  $n - k = \dim \text{ran } T$ .
  - $\dim \text{column space } A + \text{nullity } A = n$ .
  - $\text{rank } A + \text{nullity } A = n$ . The Proof is done.
- Example 1:

**Theorem 7.5.2** *If  $A$  is an  $m \times n$  matrix, then*

$$\text{rank}(A) = \text{rank}(A^T)$$

(3)

- $\text{rank}(A^T) + \text{nullity}(A^T) = m$ . ( $A^T$  is  $n \times m$  matrix)
- $\text{rank}(A) + \text{nullity}(A) = n$ .
- Thus the dimension of four fundamental space is determined from a single number rank  $A$ .
- $\dim \text{row } A = k$ ,  $\dim \text{null } A = n - k$ ,  $\dim \text{col } A = k$ ,  
 $\dim \text{null } A^T = m - k$ .
- See Example 2.

# The relationship between consistency and rank.

**Theorem 7.5.3 (The Consistency Theorem)** *If  $A\mathbf{x} = \mathbf{b}$  is a linear system of  $m$  equations in  $n$  unknowns, then the following statements are equivalent.*

- (a)  $A\mathbf{x} = \mathbf{b}$  is consistent.
- (b)  $\mathbf{b}$  is in the column space of  $A$ .
- (c) The coefficient matrix  $A$  and the augmented matrix  $[A \mid \mathbf{b}]$  have the same rank.

- Proof: (a)  $\leftrightarrow$  (b) by Theorem 3.5.5.  
(a)  $\leftrightarrow$  (c). Put both into ref. Then the number of the nonzero rows are the same for consistency.
- Example 3:

**Definition 7.5.4** An  $m \times n$  matrix  $A$  is said to have *full column rank* if its column vectors are linearly independent, and it is said to have *full row rank* if its row vectors are linearly independent.

**Theorem 7.5.5** Let  $A$  be an  $m \times n$  matrix.

- (a)  $A$  has full column rank if and only if the column vectors of  $A$  form a basis for the column space, that is, if and only if  $\text{rank}(A) = n$ .
- (b)  $A$  has full row rank if and only if the row vectors of  $A$  form a basis for the row space, that is, if and only if  $\text{rank}(A) = m$ .

- Proof: clear

**Theorem 7.5.6** *If  $A$  is an  $m \times n$  matrix, then the following statements are equivalent.*

- (a)  $A\mathbf{x} = \mathbf{0}$  has only the trivial solution.
- (b)  $A\mathbf{x} = \mathbf{b}$  has at most one solution for every  $\mathbf{b}$  in  $R^m$ .
- (c)  $A$  has full column rank.

- Proof: (a) $\leftrightarrow$ (b) Theorem 3.5.3.
- (a) $\leftrightarrow$ (c).  $A\mathbf{x} = \mathbf{0}$  can be written  $x_1\mathbf{a}_1 + \dots + x_n\mathbf{a}_n = \mathbf{0}$ .  
The trivial solution  $\leftrightarrow$   $\mathbf{a}_i$  independent.  $\leftrightarrow$   $A$  has full column rank.
- Example 5.

# Overdetermined and underdetermined

- A  $m \times n$ -matrix.
  - If  $m > n$ , then overdetermined.
  - If  $m < n$ , then underdetermined.

**Theorem 7.5.7** *Let  $A$  be an  $m \times n$  matrix.*

- (a) (**Overdetermined Case**) *If  $m > n$ , then the system  $A\mathbf{x} = \mathbf{b}$  is inconsistent for some vector  $\mathbf{b}$  in  $R^m$ .*
- (b) (**Underdetermined Case**) *If  $m < n$ , then for every vector  $\mathbf{b}$  in  $R^m$  the system  $A\mathbf{x} = \mathbf{b}$  is either inconsistent or has infinitely many solutions.*

- Proof(a):  $m > n$ . The column vectors of  $A$  cannot span  $R^m$ .
- (b):  $m < n$ . The column vectors of  $A$  is linearly dependent.  $A\mathbf{x} = \mathbf{0}$  has infinitely many solutions. Use Theorem 3.5.2.

# Matrices of form $A^T A$ and $AA^T$ .

- $AA^T$ . The  $ij$ -th entry is  $a_i \cdot a_j$ .  $a_i$  column vector
- $A^T A$ . The  $ij$ -th entry is  $r_i \cdot r_j$ .  $r_i$  row vector

**Theorem 7.5.8** *If  $A$  is an  $m \times n$  matrix, then:*

- $A$  and  $A^T A$  have the same null space.*
- $A$  and  $A^T A$  have the same row space.*
- $A^T$  and  $A^T A$  have the same column space.*
- $A$  and  $A^T A$  have the same rank.*

- Proof (a).  $\text{null } A$  is a subset of  $\text{null } A^T A$ . (if  $Ax=0$ , then  $A^T Ax=0$ ).
- $\text{null } A^T A$  is a subset of  $\text{null } A$ . (If  $A^T Av=0$ , then  $v$  is orthogonal to every row vector of  $A^T A$ . Since  $A^T A$  is symmetric,  $v$  is orthogonal to every column vectors of  $A^T A$ . Thus,  $v^T A^T Av=0$ .  $(Av)^T Av=0$ . Thus  $Av \cdot Av=0$  and  $Av=0$ ).
- (b) By Theorem 7.3.5. The complements are the same.
- (c). The column space of  $A^T$  is the row space of  $A$ .
- (d). From (b).

**Theorem 7.5.9** *If  $A$  is an  $m \times n$  matrix, then:*

- (a)  $A^T$  and  $AA^T$  have the same null space.
- (b)  $A^T$  and  $AA^T$  have the same row space.
- (c)  $A$  and  $AA^T$  have the same column space.
- (d)  $A$  and  $AA^T$  have the same rank.

# Unifying theorem.

**Theorem 7.5.10** *If  $A$  is an  $m \times n$  matrix, then the following statements are equivalent.*

- (a)  $A\mathbf{x} = \mathbf{0}$  has only the trivial solution.
- (b)  $A\mathbf{x} = \mathbf{b}$  has at most one solution for every  $\mathbf{b}$  in  $R^m$ .
- (c)  $A$  has full column rank.
- (d)  $A^T A$  is invertible.

Proof) (a) $\leftrightarrow$ (b) $\leftrightarrow$ (c). Done before.

(c) $\leftrightarrow$ (d).  $A^T A$  is an  $n \times n$  matrix.  $A^T A$  is invertible if and only if  $A^T A$  is of full rank. By Theorem 7.5.8(d), this is if and only if  $A$  is full rank.

**Theorem 7.5.11** *If  $A$  is an  $m \times n$  matrix, then the following statements are equivalent.*

- (a)  $A^T \mathbf{x} = \mathbf{0}$  has only the trivial solution.
- (b)  $A^T \mathbf{x} = \mathbf{b}$  has at most one solution for every vector  $\mathbf{b}$  in  $R^n$ .
- (c)  $A$  has full row rank.
- (d)  $AA^T$  is invertible.

- Example 7: