

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 $T: \mathbb{R}^n \rightarrow \mathbb{R}^m$ be defined by $T(x) = Ax$. Then
 - $\dim \text{ran} T = \dim \text{column space } A$.
 - $\text{Ker } T = \text{null } A$.
 - $\dim \text{ran } T + \dim \text{ker } T = n$.
 - Choose a basis a_1, \dots, a_k in $\text{ker } T$. $\dim \text{ker } T = k$.
 - 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 \mathbb{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: