

VV285 RC Part II

Elements of Linear Algebra

“Linear Algebra!”

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We briefly return to the theory of solvability of linear systems of equations $Ax = b$. We define the **solution set**

$$\text{Sol}(A, b) = \{x \in \mathbb{R}^n : Ax = b\}.$$

If $x_0 \in \mathbb{R}^n$ satisfies

$$Ax_0 = b$$

we say that x_0 is a **particular solution** of $Ax = b$. The **associated homogeneous solution set** is

$$\text{Sol}(A, 0) = \{x \in \mathbb{R}^n : Ax = 0\} = \ker A.$$

A very important, fundamental result states:

The solution set of $Ax = b$ is the sum of the homogeneous solution set and a particular solution.

6.1. Lemma. Let $x_0 \in \mathbb{R}^n$ be a particular solution of $Ax = b$. Then

$$\text{Sol}(A, b) = \{x_0\} + \ker A = \{y \in \mathbb{R}^n : y = x_0 + x, x \in \ker A\}.$$

where the sum of sets is understood as in Definition 2.24.

6.8. **Theorem.** There exists a solution x for $Ax = b$ if and only if $\text{rank } A = \text{rank}(A \mid b)$, where

$$(A \mid b) = \begin{pmatrix} a_{11} & \dots & a_{1n} & b_1 \\ \vdots & & \vdots & \vdots \\ a_{m1} & \dots & a_{mn} & b_m \end{pmatrix} \in \text{Mat}((n+1) \times m).$$

$Ax = b$ has solution $x \in \mathbb{R}^n$

$$\Leftrightarrow b \in \text{ran } A$$

$$\Leftrightarrow b \in \text{span}\{a_1, \dots, a_n\}$$

$$\Leftrightarrow b \text{ is not independent of } a_1, \dots, a_n$$

$$\Leftrightarrow \dim \text{span}\{a_1, \dots, a_n\} = \dim \text{span}\{a_1, \dots, a_n, b\}$$

$$\Leftrightarrow \dim \text{ran } A = \dim \text{ran}(A \mid b)$$

$$\Leftrightarrow \text{rank } A = \text{rank}(A \mid b)$$

6.4. Fredholm Alternative. Let A be an $n \times n$ matrix. Then

- ▶ either $Ax = b$ has a unique solution for any $b \in \mathbb{R}^n$
- ▶ or $Ax = 0$ has a non-trivial solution.

6.7. Definition and Theorem. Let $A \in \text{Mat}(m \times n; \mathbb{F})$. Then the column rank is equal to the row rank and we define the **rank** of A by

$$\text{rank } A := \text{column rank } A = \text{row rank } A.$$

Proof.

In the assignments it will be shown that

$$\text{ran } \overline{A}^T = (\ker A)^\perp.$$

Then, using Corollary 3.24 and the dimension formula (4.3),

$$\begin{aligned} \text{row rank } A &= \text{column rank } A^T = \text{column rank } \overline{A}^T = \dim \text{ran } \overline{A}^T \\ &= \dim(\ker A)^\perp = n - \dim \ker A = \dim \text{ran } A = \text{column rank } A. \end{aligned}$$

Here we have used that complex conjugation is a linear, bijective map $\mathbb{C} \rightarrow \mathbb{C}$ if \mathbb{C} is regarded as a real vector space. □

6.5. Definition. Let $A \in \text{Mat}(m \times n; \mathbb{F})$ be a matrix with columns $a_{.j} \in \mathbb{F}^m$, $1 \leq j \leq n$, and rows $a_{i.} \in \mathbb{F}^n$, $1 \leq i \leq m$. Then we define

- ▶ the **column rank** of A to be

$$\text{column rank } A := \dim \text{span}\{a_{.1}, \dots, a_{.n}\}$$

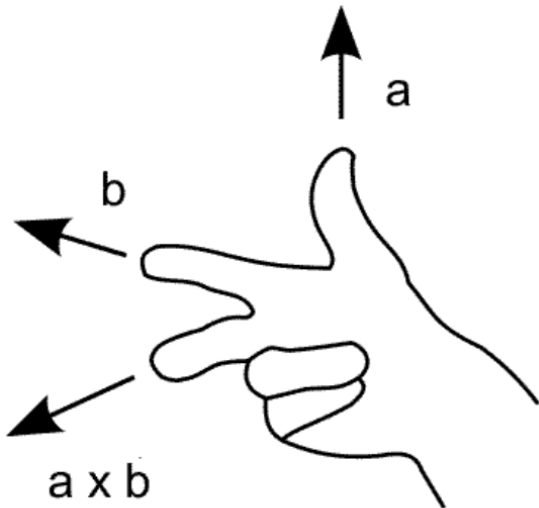
- ▶ and the **row rank** of A to be

$$\text{row rank } A := \dim \text{span}\{a_{1.}, \dots, a_{m.}\}.$$

6.6. Remarks.

- ▶ The column rank is the greatest number of independent column vectors $a_{.j}$ that can be selected from all columns. This is analogously true for the row rank.
- ▶ $\text{column rank } A = \text{row rank } A^T$.
- ▶ $\text{column rank } A = \dim \text{ran } A$.

Determinant



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