

MATH 430
Advanced Linear Algebra
Session 18

Systems of linear equations

$$A \mathbf{x} = b$$

$A \in M_{m \times n}$,

$b \in \mathbb{R}^m$.

Homogeneous if $b = \vec{0}$.

Non-homogeneous if $b \neq \vec{0}$.

Let $\varphi : \mathbb{R}^n \rightarrow \mathbb{R}^m$ be defined by $\varphi(\mathbf{x}) = A\mathbf{x}$.

$x \mapsto A\mathbf{x} = b$

Solutions of $A\mathbf{x} = 0$ form the null space of A , denoted by $N(A)$.

$x = \vec{0}$ is always a solution.
There may be other solutions.

$R(A) = \text{span of the columns of } A$.
 $b \in \text{span}\{\text{cols. of } A\}$

Recall: $A\mathbf{x} = b$ has a solution $\iff b \in \text{span}\{\text{cols. of } A\}$

max. # of l.i. columns of $A = \dim(R(A)) = \text{rank}(A)$

By the Dimension Theorem

$$\frac{\dim(\mathbb{R}^n)}{\dim(N(A))} = \dim(N(A)) + \text{rank}(A)$$

$$\dim(N(A)) = \dim(\mathbb{R}^n) - \text{rank}(A) \quad \text{Thm 3.8}$$

$$Ax = \underline{\underline{b}}$$

Let K be the solution set of $Ax = \underline{\underline{b}}$.
 Let $x = r$ be a solution. Then
 $K = \{s + r : s \in N(A)\}$.

$$s \text{ solves } Ax = \vec{0}$$

This means: any solution of $Ax = b$ can be expressed as the sum of a particular solution of $Ax = b$ and any solution of $Ax = \vec{0}$.

Example (from previous session):

$$\begin{bmatrix} 1 & 2 & 1 \\ 1 & -1 & 3 \\ 2 & 3 & 2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \\ 2 \end{bmatrix} \quad b$$

$\xrightarrow{\text{row reduction}}$

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

$\xrightarrow{\text{row reduction}}$

$$\begin{bmatrix} 1 & 2 & 1 \\ 0 & -1 & 0 \\ 0 & 0 & 0 \end{bmatrix} \quad \alpha \in \mathbb{R}$$

A

$$x_1 = 1 - \alpha, \quad x_2 = 0, \quad x_3 = \alpha, \quad \alpha \in \mathbb{R}$$

$$x = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 1 - \alpha \\ 0 \\ \alpha \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} + \alpha \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix} = k + s$$

$k = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$ (when $\alpha = 0$) solves the above system.

check that $\begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}$ solves $Ax = \vec{0}$.

$N(A) = \alpha \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix} = \text{Span} \left\{ \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix} \right\}$

$k = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$ is called a particular solution.

Unique solution : $Ax = b$ has only one solution for a given b , i.e.

$$\begin{array}{ccc} x & \longrightarrow & Ax \\ & & = b \end{array}$$

one-to-one $\iff N(A) = \{\vec{0}\} \iff Ax = \vec{0}$
has only the trivial solution $x = \vec{0}$.

In the above example, $N(A) \neq \{\vec{0}\}$
 \Rightarrow solution is not unique.

Let A be square ($n \times n$) : there are n equations in n unknowns.

$$\begin{array}{ccc} x & \xrightarrow{\quad} & A \underbrace{x}_{n \times n} = b \underbrace{\in \mathbb{R}^n}_{n \times 1} \\ \in \mathbb{R}^n & & \end{array}$$

$$\begin{array}{ccc} A : V \mathbb{R}^n & \longrightarrow & W \mathbb{R}^n \\ V = W = \mathbb{R}^n & & \end{array}$$

\Leftrightarrow A is one-one $\Leftrightarrow A$ is onto
 \Leftrightarrow cols. of A span \mathbb{R}^n .

Then A is invertible and the solution is

$$x = A^{-1} b.$$

(Review finding matrix inverses)
 $\xrightarrow{\text{reduce}} [A \mid I] \xrightarrow{\text{reduce}} [I \mid A^{-1}]$

Summary :

- (a) Solution exists $\Leftrightarrow \text{rank}(A) = \text{rank}[A|b]$
 $\Leftrightarrow b \in \text{Span}\{\text{cols. of } A\}$.
- Column space of A
- (b) $Ax = \vec{0}$ has only the trivial solution $x = \vec{0}$
 $\Leftrightarrow N(A) = \{\vec{0}\} \Leftrightarrow Ax = b$ has a unique solution.
- (c) If A is $\overset{in}{\in} M_{n \times n}$ and $N(A) = \{\vec{0}\}$, the unique solution is $x = A^{-1}b$.
- (d) If $N(A) = \{\vec{0}\}$ and $A \in M_{n \times n}$
 then $\text{rank}(A) = n$.