

Lecture VIII: §1.7 (cont.) Nonsingular matrices, §1.9 Matrix inverses & their prop.

Recall.  $\{v_1, \dots, v_p\}$  in  $\mathbb{R}^n$  are l.i. if and only if the system  $AX = \underline{0}$  ( $\Rightarrow A = [v_1 \dots v_p]$  of size  $n \times p$ ) has a unique solution (the trivial soln), i.e.  $x_1 v_1 + \dots + x_p v_p = \underline{0}$  in  $\mathbb{R}^n$  has a unique solution.

• Defined  $A$  of size  $n \times n$  is nonsingular if and only if the system  $AX = \underline{0}$  has a unique solution.

§1. Nonsingular matrices: Fix  $A$  of size  $n \times n$ .

Theorem 1:  $A$  is nonsingular if and only if its set of  $n$  columns is linearly indep. in  $\mathbb{R}^n$ .

Proof:  $AX = x_1 \text{col}_1 A + \dots + x_n \text{col}_n A = \underline{0}$  has a unique solution if and only if  $\{\text{col}_1 A, \dots, \text{col}_n A\}$  is lin. indep. in  $\mathbb{R}^n$ .  $\square$

• Why study them? They lead to compatible systems with unique solutions!

Theorem 2 Fix  $A$  of size  $n \times n$ . The equation  $A\underline{x} = \underline{b}$  has a unique solution

$\Leftrightarrow$  every  $(n \times 1)$  column vector  $\underline{b}$  if and only if  $A$  is nonsingular.

Proof ( $\Rightarrow$ ) Take  $\underline{b} = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}$  in  $\mathbb{R}^n$ . By assumption,  $AX = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}$  has a unique solution,  $\underline{x}$ . By Theorem 1, this means that  $A$  is nonsingular. So the cols of  $A$  are lin. indep.

( $\Leftarrow$ ) Pick any  $\underline{b}$  in  $\mathbb{R}^n$ , and consider the set  $\{\text{col}_1 A, \text{col}_2 A, \dots, \text{col}_n A, \underline{b}\}$  of  $n+1$  vectors in  $\mathbb{R}^n$ . Since  $n+1 > n$ , by Thm in Lecture VII we know these vectors form a lin. dependent set, hence there is a nontrivial solution  $(s_1, \dots, s_n, s_{n+1})$  to:

$$s_1 \text{col}_1 A + s_2 \text{col}_2 A + \dots + s_n \text{col}_n A + s_{n+1} \underline{b} = \underline{0} \quad (*)$$

• If  $s_{n+1} = 0$ , then  $(s_1, \dots, s_n)$  is a nontrivial solution to  $s_1 \text{col}_1 A + \dots + s_n \text{col}_n A = \underline{0}$  so  $A$  is singular. This contradicts our assumption!

We conclude from this that  $s_{n+1} \neq 0$ .

• We multiply (\*) by  $\frac{1}{s_{n+1}}$  & get

$$\underline{b} = -\frac{s_1}{s_{n+1}} \text{col}_1 A + \left(\frac{-s_2}{s_{n+1}}\right) \text{col}_2 A + \dots + \left(\frac{-s_n}{s_{n+1}}\right) \text{col}_n A.$$

$$\underline{b} = A \begin{bmatrix} -s_1/s_{n+1} \\ \vdots \\ -s_n/s_{n+1} \end{bmatrix} \quad \& \text{ so the system has a solution!}$$

$$\underline{x} = \left( \frac{-s_1}{s_{n+1}}, \dots, \frac{-s_n}{s_{n+1}} \right)$$

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• Claim: The solution is unique:

Why? Assume the vectors  $s, s'$  in  $\mathbb{R}^n$  are solutions of  $Ax=b$ . We want to show  $s=s'$ .

Indeed,  $As = As' = b$ , so we take  $0 = b - b = As - As' = A(s - s')$

So  $s - s'$  is a vector in  $\mathbb{R}^n$  solving  $Ax = 0$ .

But  $A$  is nonsingular, so the system  $\rightarrow$  has a unique solution (the trivial one).

Then  $s - s' = 0$ , which means  $s = s'$ . We conclude  $Ax=b$  has a unique solution  $\square$ .

### §1.9 Matrix Inverses and their properties:

Recall:  $a \neq 0$  in  $\mathbb{R} \implies b = \frac{1}{a}$  satisfies  $ab = ba = 1$ , and  $1a = a1 = a$

"Every nonzero real number is invertible"

( $b =$  multiplicative inverse of  $a$ )

( $1 =$  Neutral element for multiplication)

What an analog for matrices!

• We have: if  $A$  of size  $n \times n$ , then  $I_n = (n \times n)$  identity matrix satisfies

$$AI_n = I_n A = A \implies I_n = \text{neutral element for matrix multiplication}$$

• What about inverses? Not always here then

Def:  $A$  of size  $n \times n$ . We say  $A$  is invertible if we can find a matrix  $B$  of size  $n \times n$  satisfying  $AB = BA = I_n$ .

Lemma: If it exists,  $B$  is unique so we can write it as  $A^{-1}$ .

Proof: Assume we have 2 such matrices:  $B$  &  $B'$ . Then:

$$B = I_n B = (B'A)B = B'(AB) = B'I_n = B' \quad \square$$

Assoc.

Warning: Not every nonzero square matrix has an inverse!

Example:  $A = \begin{pmatrix} 1 & 2 \\ 2 & 4 \end{pmatrix}$  is NOT invertible.

Why? Write  $B = \begin{pmatrix} a & b \\ c & d \end{pmatrix} \implies \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \stackrel{?}{=} AB = \begin{pmatrix} 1 & 2 \\ 2 & 4 \end{pmatrix} \begin{pmatrix} a & b \\ c & d \end{pmatrix} = \begin{pmatrix} a+2c & b+2d \\ 2a+4c & 2b+4d \end{pmatrix}$

So  $\begin{cases} 1 = a+2c \\ 0 = 2a+4c = 2(a+2c) = 2 \cdot 1 = 2 \end{cases}$  so  $0=2$ , a contradiction!

We conclude no  $B$  satisfies  $\begin{pmatrix} 1 & 2 \\ 2 & 4 \end{pmatrix} B = I_2$ . | Note:  $\text{col}_2 A = 2 \text{col}_1 A$  so  $A$  is singular

• The example hints how to (i) decide if  $A$  is invertible or not  
(ii) construct  $A^{-1}$ , whenever it exists.

# §1. Existence of Inverses & Computation:

Theorem 3: A of size  $n \times n$  is invertible if and only if A is nonsingular

Proof: Next time! (Lecture IX)

Algorithm?  $AB = I_n$  translates to  $n$  systems of  $n$  equations in  $n$  unknowns.

$$A\underline{x} = \text{col}_1(I_n) = e_1, \dots, A\underline{x} = \text{col}_n(I_n) = e_n \quad (**)$$

(soln =  $\text{col}_1(B)$ )                      (soln =  $\text{col}_n(B)$ )

We can solve these  $n$  systems all at once!  $[A | e_1 | e_2 | \dots | e_n] = [A | I_n]$

We use Gauss-Jordan to put this augmented matrix in reduced echelon form.

$\uparrow \quad \uparrow \quad \uparrow$   
constants for each system

Claim: A invertible, so  $[A | I_n] \xrightarrow{\text{row eq}} [A' | C]$  in REF &  $A'$  cannot have a row of 0's (otherwise the systems  $(**)$  wouldn't have a unique solution!)

Since  $A'$  has size  $n \times n$  (square!) & is in REF, we have no other option but to get  $A' = I_n$ .

We conclude  $[A | I_n] \xrightarrow{\text{row eq}} [I_n | C]$  so  $AC = I_n$ . (because each one of the columns of  $C$  solves the  $n$  systems one of  $(**)$ )

Since  $A$  is invertible,  $A^{-1}(AB = AC)$  gives  $A^{-1}(AB) = A^{-1}(AC)$   
 $B = I_n B = (A^{-1}A)B \quad (A^{-1}A)C = C$

So  $B = C$  is the inverse of  $A$ .

Example:  $A = \begin{bmatrix} 0 & 0 & 1 \\ 0 & 1 & 2 \\ 3 & 4 & 1 \end{bmatrix}$  Decide if  $A$  is invertible & if so, find  $A^{-1}$ .

- cols of  $A$  are l.i.  $\Rightarrow A$  nonsingular  $\Rightarrow$  invertible.
- Use algorithm to build  $A^{-1}$ :

$$[A | I_3] = \left[ \begin{array}{ccc|ccc} 0 & 0 & 1 & 1 & 0 & 0 \\ 0 & 1 & 2 & 0 & 1 & 0 \\ 3 & 4 & 1 & 0 & 0 & 1 \end{array} \right] \xrightarrow{R_3 \leftrightarrow R_1} \left[ \begin{array}{ccc|ccc} 3 & 4 & 1 & 0 & 0 & 1 \\ 0 & 1 & 2 & 0 & 1 & 0 \\ 0 & 0 & 1 & 1 & 0 & 0 \end{array} \right] \xrightarrow{\begin{array}{l} R_2 \rightarrow R_2 - 2R_3 \\ R_1 \rightarrow R_3 - 3R_3 \end{array}}$$

$$\xrightarrow{R_1 \rightarrow R_1 - 4R_2} \left[ \begin{array}{ccc|ccc} 1 & 0 & 0 & 5 & -4 & 1 \\ 0 & 1 & 0 & -2 & 1 & 0 \\ 0 & 0 & 1 & 1 & 0 & 0 \end{array} \right]$$

$I_3$                       " $A^{-1}$ "

Check:  $\begin{bmatrix} 0 & 0 & 1 \\ 0 & 1 & 2 \\ 3 & 4 & 1 \end{bmatrix} \begin{bmatrix} 5 & -4 & 1 \\ -2 & 1 & 0 \\ 1 & 0 & 0 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$  ✓

(Next time we'll see that the other product (in the other order) has to give  $I_3$  as well, so only need to check one product!)