

Lecture 18: Matrices: Inverses and rank.

①

Last time: An $n \times n$ matrix A is invertible if there exists an $n \times n$ matrix B with $AB = BA = I_n$.

When A is invertible, the inverse is unique and written as A^{-1} .

Thm: Suppose $A \in M_{n \times n}(\mathbb{R})$. The following are equivalent: ① A is invertible

② The linear transformation $L_A: \mathbb{R}^n \rightarrow \mathbb{R}^n$
 $x \mapsto Ax$
is invertible.

③ The nullspace $\mathcal{N}(A) = \{x \in \mathbb{R}^n \mid Ax = 0\}$
is $\{0\}$.

Proof: First, we show $① \iff ②$. If ①,

$$\text{then } L_A \circ L_{A^{-1}} = L_{AA^{-1}} = L_{I_n} = I_{\mathbb{R}^n}$$

as well as $L_{A^{-1}} \circ L_A = L_{A^{-1}A} = L_{I_n} = I_{\mathbb{R}^n}$. (2)

Thus $L_{A^{-1}}$ is the inverse to L_A and so L_A is invertible. Conversely, assume (2). Let

$T: \mathbb{R}^n \rightarrow \mathbb{R}^n$ be the inverse of L_A . Set

$\beta = \{\text{std basis for } \mathbb{R}^n\}$ and $B = [T]_{\beta}$; then

$T = L_B$ since $[L_B]_{\beta} = B = [T]_{\beta}$. So

$L_{AB} = L_A \circ L_B = L_A \circ T = I_{\mathbb{R}^n}$, which gives

$AB = I_n$. Similarly $L_{BA} = L_B \circ L_A = T \circ L_A$

$= I_{\mathbb{R}^n} \Rightarrow BA = I_n$. So A is indeed

invertible with inverse B . So we've shown

(1) \Leftrightarrow (2).

Next we show (1) \Rightarrow (3). If A is invertible and $Ax = 0$, then

$$A^{-1}(Ax) = A^{-1}0 = 0$$

③

$$\text{As } A^{-1}(Ax) = (A^{-1}A)x = I_n x = x,$$

this gives $x = 0$. Thus $\mathcal{N}(A) = \{0\}$.

Finally, assume ③. As

$$\mathcal{N}(L_A) = \{x \in \mathbb{R}^n \mid \underbrace{L_A(x)}_{Ax} = 0\} = \mathcal{N}(A)$$

we get $\mathcal{N}(L_A) = 0$ which forces L_A to be 1-1, and hence onto by Theorems

2.4 and 2.5 of our text (which you used on the last HW). Thus L_A is

invertible and so ③ \Rightarrow ② \Rightarrow ① ▣

How do we compute A^{-1} ? Set $A^{-1} = B$

$$= \begin{pmatrix} | & & | \\ b_1 & \cdots & b_n \\ | & & | \end{pmatrix}. \text{ Since } AB = \begin{pmatrix} | & & | \\ Ab_1 & \cdots & Ab_n \\ | & & | \end{pmatrix} = I_n,$$

we get $Ab_i = e_i$ for each i .

(4)

Thus, we can find b_i by solving the linear system $LS(A, e_i)$. To solve all n systems at once, use this trick:

Ex: Find inverse of $A = \begin{pmatrix} 4 & 16 & 5 \\ 6 & 25 & 8 \\ 1 & 3 & 1 \end{pmatrix}$.

Consider the "super augmented" matrix

$$\left(\begin{array}{ccc|ccc} 4 & 16 & 5 & 1 & 0 & 0 \\ 6 & 25 & 8 & 0 & 1 & 0 \\ 1 & 3 & 1 & 0 & 0 & 1 \end{array} \right)$$

Now do row ops to put this in reduced row echelon form. One gets

$$\left(\begin{array}{ccc|ccc} 1 & 0 & 0 & 1 & -1 & 3 \\ 0 & 1 & 0 & 2 & -1 & -2 \\ 0 & 0 & 1 & -7 & 4 & 4 \end{array} \right)$$

\swarrow b_1 \swarrow b_2 \swarrow b_3

and so $A^{-1} = \begin{pmatrix} 1 & -1 & 3 \\ 2 & -1 & -2 \\ -7 & 4 & 4 \end{pmatrix}$.

Note: if A is not invertible, will end up with left-hand part $\neq I_n$ and

one of the systems being inconsistent.

(5)

Application: When know A^{-1} can use it to solve any LS(A, b) since $Ax = b \Rightarrow x = A^{-1}b$.

Suppose $A \in M_{m \times n}(\mathbb{R})$. We can associate

3 subspaces to A: [The 1st two we've seen before.]

$$\mathcal{N}(A) = \{x \in \mathbb{R}^n \mid Ax = 0\}$$

$$\text{RowSp}(A) = \text{span}(\text{rows of } A) \subseteq \mathbb{R}^n$$

$$\text{ColSp}(A) = \text{span}(\text{cols of } A) \subseteq \mathbb{R}^m$$

In terms of $L_A: \mathbb{R}^n \rightarrow \mathbb{R}^m$ have.

$$x \longmapsto Ax$$

$$\mathcal{N}(A) = \mathcal{N}(L_A) \text{ and } \text{ColSp}(A) = \mathcal{R}(L_A)$$

where the latter is because $\mathcal{R}(L_A)$ is spanned by the the image of any basis.

By the Dimension Thm applied to LA

(6)

we have

$$\dim \mathcal{N}(A) + \dim \text{ColSp}(A) = \dim \mathbb{R}^n = n.$$

Now, back in Lecture 12, we saw

$$\dim \mathcal{N}(A) + \dim \text{RowSp}(A) = n$$

by using that for matrices B in reduced row echelon form one has

$$\dim \mathcal{N}(B) = \# \text{ of non-pivot cols}$$

$$\begin{aligned} \dim \text{RowSp}(B) &= \# \text{ of nonzero rows} \\ &= \# \text{ of pivot cols.} \end{aligned}$$

[Recall that row equivalent matrices have the same null space and row space.]

Thus we get

Cor: $\dim \text{RowSp}(A) = \dim \text{ColSp}(A).$

This number is usually called the rank of A .