

9/17/2021

last time:

→ Example of Gaussian elimination

→ matrix inverses.

Def: The row echelon form of a matrix $\underline{A}$, denoted $\text{ref}(\underline{A})$, is the matrix $\underline{U}$ that is the result of applying Gaussian elimination to $\underline{A}$. If $\underline{A}$ is square, $\underline{U}$ is upper triangular.

Theorem: (about matrix inverses)

let $\underline{A} \in \mathbb{R}^{n \times n}$

let $\text{ref}(\underline{A}) = \underline{U}$ ← matrix obtained after Gaussian elimination

The following statements are equivalent

(a) (left inverse) There exists a matrix $\underline{Y} \in \mathbb{R}^{n \times n}$ such that

$$\underline{Y} \underline{A} = \underline{I}.$$

(b) $\underline{A} \underline{x} = \underline{0}$ implies $\underline{x} = \underline{0}$.

(c) $\text{ref}(\underline{A})$ has no zeros on the diagonal.

(d) $\underline{A} \underline{x} = \underline{b}$ has a solution for all $\underline{b}$.

(e) ~~(left inverse)~~ (right inverse) There exists a matrix $\underline{X} \in \mathbb{R}^{n \times n}$ such that

$$\underline{A} \underline{X} = \underline{I}.$$

Also, if $\underline{X}$ and $\underline{Y}$ above exist,

then $\underline{X} = \underline{Y}$.

(left and right inverses are equal).

(a) $\Rightarrow$ (b): Assume $\underline{Y}$ exists, $\underline{Y} \underline{A} = \underline{I}$.

Given $\underline{A} \underline{x} = \underline{0}$, multiply on the left by $\underline{Y}$: $\underline{Y} \underline{A} \underline{x} = \underline{I} \underline{x} = \underline{x} = \underline{0}$.

(b) $\Rightarrow$ (c): We show the contrapositive i.e., assume $\text{ref}(\underline{A}) = \underline{U}$ has at least one zero on the diagonal. Then, there would be some $\underline{x}^* \neq \underline{0}$ such that

$$\underline{U} \underline{x}^* = \underline{0}, \text{ implying } \underline{A} \underline{x}^* = \underline{0}.$$

(c) $\Rightarrow$ (d): If $\text{ref}(\underline{A}) = \underline{U}$ has no zeros on the diagonal, then a solution to $\underline{A} \underline{x} = \underline{b}$, for any $\underline{b}$, can be computed by Gaussian elimination + back substitution.

(d) $\Rightarrow$ (e). Assume $\underline{A} \underline{x} = \underline{b}$ has a solution for any $\underline{b}$. Define the columns of

$$\underline{X} = \begin{pmatrix} | & & | \\ \underline{x}_1 & \dots & \underline{x}_n \\ | & & | \end{pmatrix}$$

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as solutions to $\underline{A} \underline{x}_j = \underline{e}_j$.

Then, by construction,

$$\underline{A} \underline{X} = \begin{pmatrix} \underline{A} \underline{x}_1 & \cdots & \underline{A} \underline{x}_n \\ | & & | \end{pmatrix} = \begin{pmatrix} \underline{e}_1 & \cdots & \underline{e}_n \\ | & & | \end{pmatrix} = \underline{I}.$$

(c) $\Rightarrow$ (a): Assume we have $\underline{X}$ satisfying

$$\underline{A} \underline{X} = \underline{I}. \quad \text{So, } \underline{A} \text{ is a left inverse}$$

for $\underline{X}$. From (a) $\Rightarrow$ (c), $\underline{X}$ has

a right inverse, call it $\underline{B}$:

$$\underline{X} \underline{B} = \underline{I}.$$

$$\text{Now, we have: } \underline{A} = \underline{A} \underline{I} = \underline{A} \underline{X} \underline{B} = \underline{I} \underline{B} = \underline{B}.$$

Since $\underline{A} = \underline{B}$ and $\underline{X} \underline{B} = \underline{I}$, we have

shown that $\underline{X}$ is also a left inverse

for $\underline{A}$.

Now, to show that left and right inverses are equal...

Assume we have $\underline{X}$, $\underline{Y}$ so that

$$\underline{Y} \underline{A} = \underline{I} = \underline{A} \underline{X}.$$

Then, $\underline{Y} = \underline{Y} \underline{I} = \underline{Y} \underline{A} \underline{X} = \underline{I} \underline{X} = \underline{X}.$

Question: How could we compute $\underline{A}^{-1}$?

using Gaussian elimination:

In our proof above, we saw that by solving

$$\underline{A} \underline{x} = \underline{e}_j$$

the solution $\underline{x}$ is the j th column of $\underline{A}^{-1}$.

We can solve the system above by constructing the augmented matrix

$$(\underline{A} \mid \underline{e}_j)$$

$\leadsto$

As before, Gaussian elimination gives $(\underline{U} \mid \underline{c})$, where $\underline{U}$ is upper triangular. If we perform further row operations to transform $\underline{U}$ to the identity matrix, then we have solved our system:

$$(\underline{A} \mid \underline{e}_j) \xrightarrow{\text{Gaussian elimination}} (\underline{U} \mid \underline{c}) \xrightarrow{\text{more row operations}} (\underline{I} \mid \underline{x})$$

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 Gaussian elimination more row operations solution

where, again, $\underline{x}$ is the j th column of $\underline{A}^{-1}$

We can compute all the columns of $\underline{A}^{-1}$ at once by doing this process on

$\underline{I}$

$$\left(\underline{\underline{A}} \mid \begin{matrix} \underline{\underline{e}}_1 \\ \underline{\underline{e}}_2 \\ \vdots \\ \underline{\underline{e}}_n \end{matrix} \right) = \left(\underline{\underline{A}} \mid \underline{\underline{I}} \right),$$

i.e. we do:

$$\left(\underline{\underline{A}} \mid \underline{\underline{I}} \right) \xrightarrow[\text{row operations.}]{\quad} \left(\underline{\underline{I}} \mid \underline{\underline{A}}^{-1} \right)$$

Ex: compute the inverse of

$$\underline{\underline{A}} = \begin{pmatrix} 2 & 4 & -2 \\ 4 & 9 & -3 \\ -2 & -3 & 7 \end{pmatrix}$$

setup the augmented matrix:

$$\left(\underline{\underline{A}} \mid \underline{\underline{I}} \right) = \left(\begin{array}{ccc|ccc} 2 & 4 & -2 & 1 & 0 & 0 \\ 4 & 9 & -3 & 0 & 1 & 0 \\ -2 & -3 & 7 & 0 & 0 & 1 \end{array} \right).$$

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Now we do row operations to put A in upper triangular form...

$$\left(\begin{array}{ccc|ccc} 2 & 4 & -2 & 1 & 0 & 0 \\ 4 & 9 & -3 & 0 & 1 & 0 \\ -2 & -3 & 7 & 0 & 0 & 1 \end{array} \right)$$



$$\left(\begin{array}{ccc|ccc} 2 & 4 & -2 & 1 & 0 & 0 \\ 4 & 9 & -3 & 0 & 1 & 0 \\ 0 & 1 & 5 & 1 & 0 & 1 \end{array} \right)$$



$$\left(\begin{array}{ccc|ccc} 2 & 4 & -2 & 1 & 0 & 0 \\ 0 & 1 & 1 & -2 & 1 & 0 \\ 0 & 1 & 5 & 1 & 0 & 1 \end{array} \right)$$



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

Now we do row operations to transform the matrix on the left to a diagonal matrix



$$\left(\begin{array}{ccc|ccc} 2 & 4 & 0 & 5/2 & -1/2 & 1/2 \\ 0 & 1 & 1 & -2 & 1 & 0 \\ 0 & 0 & 4 & 3 & -1 & 1 \end{array} \right)$$

$$\downarrow$$

$$\left(\begin{array}{ccc|ccc} 2 & 4 & 0 & 5/2 & -1/2 & 1/2 \\ 0 & 1 & 0 & -11/4 & 5/4 & -1/4 \\ 0 & 0 & 4 & 3 & -1 & 1 \end{array} \right)$$

$$\downarrow$$

$$\left(\begin{array}{ccc|ccc} 2 & 0 & 0 & 27/2 & -11/2 & 3/2 \\ 0 & 1 & 0 & -11/4 & 5/4 & -1/4 \\ 0 & 0 & 4 & 3 & -1 & 1 \end{array} \right)$$

Now, we divide each row so I appears on the left:

$$\left(\begin{array}{ccc|ccc} 1 & 0 & 0 & 27/4 & -11/4 & 3/4 \\ 0 & 1 & 0 & -11/4 & 5/4 & -1/4 \\ 0 & 0 & 1 & 3/4 & -1/4 & 1/4 \end{array} \right),$$

$$\Rightarrow \underline{\underline{A^{-1}}} = \frac{1}{4} \left(\begin{array}{ccc} 27 & -11 & 3 \\ -11 & 5 & -1 \\ 3 & -1 & 1 \end{array} \right).$$