

LAST TIME

- * Matrix operations
- * Interpretations of matrix multiplication
- * Block matrices

TODAY

- * Inverse Matrices

THE INVERSE

Recall: $B \in \mathbb{R}^{n \times n}$ is the inverse of $A \in \mathbb{R}^{n \times n}$ if

$$BA \vec{x} = \vec{x} \quad \text{for all } \vec{x} \in \mathbb{R}^n$$

$= \underline{I} \vec{x}$ where \underline{I} is the $n \times n$ identity matrix

$$\underline{I} = \begin{bmatrix} 1 & 0 & 0 & \dots & 0 \\ 0 & 1 & 0 & & 0 \\ & 0 & 1 & & 0 \\ \vdots & & & \ddots & \vdots \\ 0 & & & & 1 \end{bmatrix}$$

Hence, $B \in \mathbb{R}^{n \times n}$ is the inverse of $A \in \mathbb{R}^{n \times n}$ if and only if $BA = \underline{I}$.

Note: Although $BA \neq AB$ in general, if $BA = \underline{I}$, then $AB = \underline{I}$ too!

Uniqueness of the Inverse

Can $A \in \mathbb{R}^{n \times n}$ have two inverses? No

Why? If $BA = I = CA$,
then $AB = I$ too.

So, $CA = I$

$\Rightarrow (CA)B = B$ (multiplying by B from the right)

$\Rightarrow C(AB) = B$ (associative law)

$\Rightarrow C = B$. (since $AB = I$)

A can have only one inverse and we will denote it as A^{-1}

Notes on Inverses

Note 1

The inverse exists if and only if elimination produces n pivots

Note 2

If A is invertible, the one and only solution to $A\vec{x} = \vec{b}$ is $\vec{x} = \bar{A}^{-1}\vec{b}$

Why? $A\vec{x} = \vec{b} \xrightarrow[\text{by } \bar{A}^{-1}]{\text{multiply}} \bar{A}^{-1}(A\vec{x}) = \bar{A}^{-1}\vec{b}$ or $(\bar{A}^{-1}A)\vec{x} = \bar{A}^{-1}\vec{b}$

Note 3

Suppose there is a non-zero vector \vec{x} such that $A\vec{x} = \vec{0}$.

Then A cannot have an inverse.

Note 4

If A is diagonal, $A = \begin{bmatrix} d_1 & & & \\ & d_2 & & \\ & & \dots & \\ & & & d_n \end{bmatrix}$, then $\bar{A}^{-1} = \begin{bmatrix} 1/d_1 & & & \\ & 1/d_2 & & \\ & & \dots & \\ & & & 1/d_n \end{bmatrix}$

Note 5

If A and B are invertible, then $(AB)^{-1} = \bar{B}^{-1}\bar{A}^{-1}$

Why? $(AB)(\bar{B}^{-1}\bar{A}^{-1}) = A(B\bar{B}^{-1})\bar{A}^{-1} = AIA^{-1} = I$

FINDING THE INVERSE

We know that $\vec{A}^{-1}A = A\vec{A}^{-1} = I$

Hence
$$A \underbrace{\begin{bmatrix} | & | & & | \\ \vec{a}_1^{-1} & \vec{a}_2^{-1} & \dots & \vec{a}_n^{-1} \\ | & | & & | \end{bmatrix}}_{\substack{\text{column vectors} \\ \text{of } \vec{A}^{-1}}} = I$$

Therefore
$$\begin{bmatrix} | & | & & | \\ A\vec{a}_1^{-1} & A\vec{a}_2^{-1} & \dots & A\vec{a}_n^{-1} \\ | & | & & | \end{bmatrix} = \begin{bmatrix} 1 & 0 & & 0 \\ 0 & 1 & & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & & 1 \end{bmatrix}$$

This is equivalent to the n matrix-vector equations

$$A\vec{a}_1^{-1} = \begin{pmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix}, \quad A\vec{a}_2^{-1} = \begin{pmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{pmatrix}, \quad \dots, \quad A\vec{a}_n^{-1} = \begin{pmatrix} 0 \\ \vdots \\ 0 \\ 1 \end{pmatrix}$$

columns of I

Defⁿ

Let $\vec{e}_j \in \mathbb{R}^n$ be the vector

$$\begin{pmatrix} 0 \\ \vdots \\ 0 \\ 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix}$$

j^{th} entry is a 1,
the rest are 0

$\bar{A}^{-1}A = I$ if and only if $A \vec{a}_j^{-1} = \vec{e}_j$ for all $j=1, 2, \dots, n$

j^{th} column of \bar{A}^{-1}

\hookrightarrow Therefore, we can solve for \bar{A}^{-1} by performing elimination n times

$$A \vec{x} = \vec{e}_j \text{ has solution } \vec{x} = \vec{a}_j^{-1}$$

$[A \mid \vec{e}_j]$ $\xrightarrow{\text{make upper triangular and back substitute}}$ $\left. \vphantom{[A \mid \vec{e}_j]} \right\} n \text{ times}$

Fast Method When making the augmented matrix upper triangular, our work only depends on A .

Therefore, we can solve for all the equations at once!

Example: Find the inverse of $A = \begin{bmatrix} 1 & 2 & 0 \\ 2 & 2 & 1 \\ 0 & 2 & -2 \end{bmatrix}$ using elimination

Step 1 Construct augmented matrix

$[A \ I]$

$$\left[\begin{array}{ccc|ccc} 1 & 2 & 0 & 1 & 0 & 0 \\ 2 & 2 & 1 & 0 & 1 & 0 \\ 0 & 2 & -2 & 0 & 0 & 1 \end{array} \right]$$

$\uparrow \quad \uparrow \quad \uparrow$
 $\vec{e}_1 \quad \vec{e}_2 \quad \vec{e}_3$

Step 2

Perform
elimination

$$\left[\begin{array}{ccc|ccc} 1 & 2 & 0 & 1 & 0 & 0 \\ 2 & 2 & 1 & 0 & 1 & 0 \\ 0 & 2 & -2 & 0 & 0 & 1 \end{array} \right]$$

Gaussian
elimination

→

$$\left[\begin{array}{ccc|ccc} 1 & 2 & 0 & 1 & 0 & 0 \\ 0 & -2 & 1 & -2 & 1 & 0 \\ 0 & 2 & -2 & 0 & 0 & 1 \end{array} \right]$$

row ② - 2row ①

finish
with
back
substitution

→

$$\left[\begin{array}{ccc|ccc} 1 & 2 & 0 & 1 & 0 & 0 \\ 0 & -2 & 1 & -2 & 1 & 0 \\ 0 & 0 & -1 & -2 & 1 & 1 \end{array} \right]$$

row ③ + row ②

Step 3

Perform back
Substitution
bottom to top; want
zeros above pivots too

$$\left[\begin{array}{ccc|ccc} 1 & 2 & 0 & 1 & 0 & 0 \\ 0 & -2 & 1 & -2 & 1 & 0 \\ 0 & 0 & -1 & -2 & 1 & 1 \end{array} \right]$$

Gauss

- Jordan

Method

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

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

Reduced
echelon form

$$\left[\begin{array}{ccc|ccc} 1 & 0 & 0 & -3 & 2 & 1 \\ 0 & 1 & 0 & 2 & -1 & -\frac{1}{2} \\ 0 & 0 & 1 & 2 & -1 & -1 \end{array} \right] \text{Divide each row by pivot}$$

A^{-1}

Another Interpretation

$$\underbrace{[A \mid I]}_{\text{augmented matrix}} \xrightarrow{\text{elimination}} D^{-1} \underbrace{E_{q_1} E_{q_2} \dots P \dots E_1}_{\text{elimination and/or row exchange matrices}} [A \mid I] = [I \mid A^{-1}]$$

Gauss Jordan

If $E = D^{-1} E_{q_1} E_{q_2} \dots P \dots E_1$, then $E[A \mid I] = [EA \mid EI] = [I \mid A^{-1}]$

Therefore A^{-1} is a product of swap/elimination matrices

Singular vs Invertible

A^{-1} exists (and Gauss-Jordan finds it) exactly when A has n pivots

n pivots — invertible
 $< n$ pivots — singular