

Section 2.2: Inverse of a Matrix

If A is an $n \times n$ matrix, we seek a matrix A^{-1} that satisfies the condition

$$A^{-1}A = AA^{-1} = I_n.$$

If such matrix A^{-1} exists, we'll say that A is **nonsingular** (a.k.a. *invertible*). Otherwise, we'll say that A is **singular**.

Theorem (2×2 case)

Let $A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$. If $ad - bc \neq 0$, then A is invertible and

$$A^{-1} = \frac{1}{ad - bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}.$$

If $ad - bc = 0$, then A is singular.

The quantity $ad - bc$ is called the **determinant** of A .

Theorem: If A is an invertible $n \times n$ matrix, then for each \mathbf{b} in \mathbb{R}^n , the equation $A\mathbf{x} = \mathbf{b}$ has unique solution $\mathbf{x} = A^{-1}\mathbf{b}$.

Example

Solve the system using a matrix inverse

$$\begin{array}{rcrcrcrcl} 4x_1 & + & x_2 & = & 7 \\ -2x_1 & + & 3x_2 & = & 7 \end{array}$$

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

$$A = \begin{bmatrix} 4 & 1 \\ -2 & 3 \end{bmatrix} \quad \vec{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \quad \vec{b} = \begin{bmatrix} 7 \\ 7 \end{bmatrix}$$

$$\det(A) = 4 \cdot 3 - (-2)(1) = 14 \neq 0$$

A^{-1} exists and

$$A^{-1} = \frac{1}{14} \begin{bmatrix} 3 & -1 \\ 2 & 4 \end{bmatrix} = \begin{bmatrix} \frac{3}{14} & -\frac{1}{14} \\ \frac{2}{14} & \frac{4}{14} \end{bmatrix}$$

The solution

$$\vec{x} = A^{-1} \vec{b} = \frac{1}{14} \begin{bmatrix} 3 & -1 \\ 2 & 4 \end{bmatrix} \begin{bmatrix} 7 \\ 7 \end{bmatrix}$$

$$= \frac{1}{14} \begin{bmatrix} 14 \\ 42 \end{bmatrix} = \begin{bmatrix} 1 \\ 3 \end{bmatrix}$$

Theorem

(i) If A is invertible, then A^{-1} is also invertible and

$$(A^{-1})^{-1} = A.$$

(ii) If A and B are invertible $n \times n$ matrices, then the product AB is also invertible¹ with

$$(AB)^{-1} = B^{-1}A^{-1}.$$

(iii) If A is invertible, then so is A^T . Moreover

$$(A^T)^{-1} = (A^{-1})^T.$$

¹This can generalize to the product of k invertible matrices. 

Elementary Matrices

Definition: An **elementary** matrix is a square matrix obtained from the identity by performing one elementary row operation.

Examples:

$$E_1 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 3 & 0 \\ 0 & 0 & 1 \end{bmatrix}, \quad E_2 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 2 & 0 & 1 \end{bmatrix}, \quad E_3 = \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}.$$

$$3R_2 \rightarrow R_2$$

$$2R_1 + R_3 \rightarrow R_3$$

$$R_1 \leftrightarrow R_2$$

Action of Elementary Matrices

Let $A = \begin{bmatrix} a & b & c \\ d & e & f \\ g & h & i \end{bmatrix}$, and compute the following products

E_1A , E_2A , and E_3A .

$$E_1A = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 3 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} a & b & c \\ d & e & f \\ g & h & i \end{bmatrix} = \begin{bmatrix} a & b & c \\ 3d & 3e & 3f \\ g & h & i \end{bmatrix}$$

$3R_2 \rightarrow R_2$

$$E_1 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 3 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

$$E_2 A = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 2 & 0 & 1 \end{bmatrix} \begin{bmatrix} a & b & c \\ d & e & f \\ g & h & i \end{bmatrix}$$

$$= \begin{bmatrix} a & b & c \\ d & e & f \\ 2a+g & 2b+h & 2c+i \end{bmatrix}$$

$$2R_1 + R_3 \rightarrow R_3$$

$$E_2 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 2 & 0 & 1 \end{bmatrix}$$

$$E_3 A = \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} a & b & c \\ d & e & f \\ g & h & i \end{bmatrix}$$

$$= \begin{bmatrix} d & e & f \\ a & b & c \\ g & h & i \end{bmatrix}$$

$$R_1 \leftrightarrow R_2$$

$$E_3 = \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Remarks

- ▶ Elementary row operations can be equated with matrix multiplication (multiply on the left by an elementary matrix),
- ▶ Each elementary matrix is invertible where the inverse *undoes* the row operation,
- ▶ Reduction to rref is a sequence of row operations, so it is a sequence of matrix multiplications

$$\text{rref}(A) = E_k \cdots E_2 E_1 A.$$

Theorem

An $n \times n$ matrix A is invertible if and only if it is row equivalent to the identity matrix I_n . Moreover, if

$$\text{rref}(A) = E_k \cdots E_2 E_1 A = I_n, \quad \text{then} \quad A = (E_k \cdots E_2 E_1)^{-1} I_n.$$

That is,

$$A^{-1} = \left[(E_k \cdots E_2 E_1)^{-1} \right]^{-1} = E_k \cdots E_2 E_1.$$

The sequence of operations that reduces A to I_n , transforms I_n into A^{-1} .

This last observation—operations that take A to I_n also take I_n to A^{-1} —gives us a method for computing an inverse!

Algorithm for finding A^{-1}

To find the inverse of a given matrix A :

- ▶ Form the $n \times 2n$ augmented matrix $[A \quad I]$.
- ▶ Perform whatever row operations are needed to get the first n columns (the A part) to rref.
- ▶ If $\text{rref}(A)$ is I , then $[A \quad I]$ is row equivalent to $[I \quad A^{-1}]$, and the inverse A^{-1} will be the last n columns of the reduced matrix.
- ▶ If $\text{rref}(A)$ is NOT I , then A is not invertible.

Remarks: We don't need to know ahead of time if A is invertible to use this algorithm.

If A is singular, we can stop as soon as it's clear that $\text{rref}(A) \neq I$.

Examples: Find the Inverse if Possible

(a)
$$\begin{bmatrix} 1 & 2 & -1 \\ -4 & -7 & 3 \\ -2 & -6 & 4 \end{bmatrix}$$

Set up $[A \ I]$

$$\begin{bmatrix} 1 & 2 & -1 & 1 & 0 & 0 \\ -4 & -7 & 3 & 0 & 1 & 0 \\ -2 & -6 & 4 & 0 & 0 & 1 \end{bmatrix}$$

$$4R_1 + R_2 \rightarrow R_2$$

$$2R_1 + R_3 \rightarrow R_3$$

$$\begin{array}{cccccc} 4 & 8 & -4 & 4 & 0 & 0 \\ -4 & -7 & 3 & 0 & 1 & 0 \end{array}$$

$$\begin{array}{cccccc} 2 & 4 & -2 & 2 & 0 & 0 \\ -2 & -6 & 4 & 0 & 0 & 1 \end{array}$$

$$\begin{bmatrix} 1 & 2 & -1 & 1 & 0 & 0 \\ 0 & 1 & -1 & 4 & 1 & 0 \\ 0 & -2 & 2 & 2 & 0 & 1 \end{bmatrix}$$

$$2R_2 + R_3 \rightarrow R_3$$

$$\begin{bmatrix} 1 & 2 & -1 & 1 & 0 & 0 \\ 0 & 1 & -1 & 4 & 1 & 0 \\ 0 & 0 & 0 & 10 & 2 & 1 \end{bmatrix}$$

$$\begin{array}{cccccc} 0 & 2 & -2 & 8 & 2 & 0 \\ 0 & -2 & 2 & 2 & 0 & 1 \end{array}$$

A has only 2 pivot columns so

$$\text{rref}(A) \neq I$$

A is singular.

Examples: Find the Inverse if Possible

(b) $\begin{bmatrix} 1 & 2 & 3 \\ 0 & 1 & 4 \\ 5 & 6 & 0 \end{bmatrix} \stackrel{=A}{\text{Form } [A \ I]}$

$$\begin{bmatrix} 1 & 2 & 3 & 1 & 0 & 0 \\ 0 & 1 & 4 & 0 & 1 & 0 \\ 5 & 6 & 0 & 0 & 0 & 1 \end{bmatrix}$$

$$-5R_1 + R_3 \rightarrow R_3$$

$$\begin{array}{cccccc} -5 & -10 & -15 & -5 & 0 & 0 \\ 5 & 6 & 0 & 0 & 0 & 1 \end{array}$$

$$\begin{bmatrix} 1 & 2 & 3 & 1 & 0 & 0 \\ 0 & 1 & 4 & 0 & 1 & 0 \\ 0 & -4 & -15 & -5 & 0 & 1 \end{bmatrix}$$

$$4R_2 + R_3 \rightarrow R_3$$

$$\begin{bmatrix} 1 & 2 & 3 & 1 & 0 & 0 \\ 0 & 1 & 4 & 0 & 1 & 0 \\ 0 & 0 & 1 & -5 & 4 & 1 \end{bmatrix}$$

$$\begin{bmatrix} 1 & 2 & 0 & 16 & -12 & -3 \\ 0 & 1 & 0 & 20 & -15 & -4 \\ 0 & 0 & 1 & -5 & 4 & 1 \end{bmatrix}$$

Finally $-2R_2 + R_1 \rightarrow R_1$

$$\begin{bmatrix} 0 & 4 & 16 & 0 & 4 & 0 \\ 0 & -4 & -15 & -5 & 0 & 1 \end{bmatrix}$$

$$-4R_3 + R_2 \rightarrow R_2$$

$$-3R_3 + R_1 \rightarrow R_1$$

$$\begin{bmatrix} 0 & 0 & -4 & 20 & -16 & -4 \\ 0 & 1 & 4 & 0 & 1 & 0 \end{bmatrix}$$

$$\begin{bmatrix} 0 & 0 & -3 & 15 & -12 & -3 \\ 1 & 2 & 3 & 1 & 0 & 0 \end{bmatrix}$$

$$\begin{bmatrix} 1 & 0 & 0 & -24 & 18 & 5 \\ 0 & 1 & 0 & 20 & -15 & -4 \\ 0 & 0 & 1 & -5 & 4 & 1 \end{bmatrix}$$

$$\begin{array}{cccccc} 0 & -2 & 0 & -40 & 30 & 8 \\ 1 & 2 & 0 & 16 & -12 & -3 \end{array}$$

$$A^{-1} = \begin{bmatrix} -24 & 18 & 5 \\ 20 & -15 & -4 \\ -5 & 4 & 1 \end{bmatrix}$$

Solve the linear system if possible

$$x_1 + 2x_2 + 3x_3 = 3$$

$$x_2 + 4x_3 = 3$$

$$5x_1 + 6x_2 = 4$$

The coefficient matrix $A = \begin{bmatrix} 1 & 2 & 3 \\ 0 & 1 & 4 \\ 5 & 6 & 0 \end{bmatrix}$

has inverse $A^{-1} = \begin{bmatrix} -24 & 18 & 5 \\ 20 & -15 & -4 \\ -5 & 4 & 1 \end{bmatrix}$

$$\vec{x} = A^{-1} \vec{b} = \begin{bmatrix} -24 & 18 & 5 \\ 20 & -15 & -4 \\ -5 & 4 & 1 \end{bmatrix} \begin{bmatrix} 3 \\ 3 \\ 4 \end{bmatrix}$$

$-24 \cdot 3 + 18 \cdot 3 + 5 \cdot 4$
 $20 \cdot 3 + (-15) \cdot 3 + (-4) \cdot 4$
 $-5 \cdot 3 + 4 \cdot 3 + 1 \cdot 4$

$$= \begin{bmatrix} 2 \\ -1 \\ 1 \end{bmatrix}$$

Section 2.3: Characterization of Invertible Matrices

Given an $n \times n$ matrix A , we can think of

- ▶ A matrix equation $A\mathbf{x} = \mathbf{b}$;
- ▶ A linear system that has A as its coefficient matrix;
- ▶ A linear transformation $T : \mathbb{R}^n \longrightarrow \mathbb{R}^n$ defined by $T(\mathbf{x}) = A\mathbf{x}$;
- ▶ Not to mention things like its **pivots**, its **rref**, the linear dependence/independence of its columns, blah blah blah...

Question: How is this stuff related, and how does being singular or invertible tie in?

Theorem: Suppose A is $n \times n$. The following are equivalent.²

- (a) A is invertible.
- (b) A is row equivalent to I_n .
- (c) A has n pivot positions.
- (d) $A\mathbf{x} = \mathbf{0}$ has only the trivial solution.
- (e) The columns of A are linearly independent.
- (f) The transformation $\mathbf{x} \mapsto A\mathbf{x}$ is one to one.
- (g) $A\mathbf{x} = \mathbf{b}$ is consistent for every \mathbf{b} in \mathbb{R}^n .
- (h) The columns of A span \mathbb{R}^n .
- (i) The transformation $\mathbf{x} \mapsto A\mathbf{x}$ is onto.
- (j) There exists an $n \times n$ matrix C such that $CA = I$.
- (k) There exists an $n \times n$ matrix D such that $AD = I$.
- (l) A^T is invertible.

²Meaning all are true or none are true.

Theorem: (An inverse matrix is unique.)

Let A and B be $n \times n$ matrices. If $AB = I$, then A and B are both invertible with $A^{-1} = B$ and $B^{-1} = A$.

Consider the homogeneous equation $B\vec{x} = \vec{0}$,

$$\vec{x} = I\vec{x} = AB\vec{x} = A(B\vec{x}) = A\vec{0} = \vec{0}.$$

So the homogeneous equation has only the trivial solution so B^{-1} exists.

Now $AB = I$ multiply on the right by B^{-1}

$$AB\bar{B}^{-1} = I\bar{B}^{-1}$$

$$A(B\bar{B}^{-1}) = \bar{B}^{-1}$$

$$AI = \bar{B}^{-1} \Rightarrow A = \bar{B}^{-1}.$$

Since \bar{B}^{-1} is invertible

$$(A)^{-1} = (\bar{B}^{-1})^{-1} \Rightarrow A^{-1} = \bar{B}.$$

Invertible Linear Transformations

Definition: A linear transformation $T : \mathbb{R}^n \longrightarrow \mathbb{R}^n$ is said to be **invertible** if there exists a function $S : \mathbb{R}^n \longrightarrow \mathbb{R}^n$ such that both

$$S(T(\mathbf{x})) = \mathbf{x} \quad \text{and} \quad T(S(\mathbf{x})) = \mathbf{x}$$

for every \mathbf{x} in \mathbb{R}^n .

If such a function exists, we typically denote it by

$$S = T^{-1}.$$

Theorem (Invertibility of a linear transformation and its matrix)

Let $T : \mathbb{R}^n \longrightarrow \mathbb{R}^n$ be a linear transformation and A its standard matrix. Then T is invertible if and only if A is invertible. Moreover, if T is invertible, then

$$T^{-1}(\mathbf{x}) = A^{-1}\mathbf{x}$$

for every \mathbf{x} in \mathbb{R}^n .

Example

Use the standard matrix to determine if the linear transformation is invertible. If it is invertible, characterize the inverse transformation.

$$T: \mathbb{R}^2 \longrightarrow \mathbb{R}^2, \quad \text{given by} \quad T(x_1, x_2) = (3x_1 - x_2, 4x_2).$$

Find the standard matrix A .

$$T(\vec{e}_1) = T(1, 0) = (3, 0), \quad T(\vec{e}_2) = T(0, 1) = (-1, 4)$$

$$A = \begin{bmatrix} 3 & -1 \\ 0 & 4 \end{bmatrix} \quad \det(A) = 3 \cdot 4 - 0 \cdot (-1) = 12 \neq 0$$

A is invertible

$$A^{-1} = \frac{1}{12} \begin{bmatrix} 4 & 1 \\ 0 & 3 \end{bmatrix} = \begin{bmatrix} \frac{1}{3} & \frac{1}{12} \\ 0 & \frac{1}{4} \end{bmatrix}$$

$$T^{-1} \text{ exists and } T^{-1}(\vec{x}) = \vec{A}^{-1} \vec{x}$$

$$\vec{A}^{-1} \vec{x} = \begin{bmatrix} \frac{1}{3} & \frac{1}{12} \\ 0 & \frac{1}{4} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} \frac{1}{3}x_1 + \frac{1}{12}x_2 \\ \frac{1}{4}x_2 \end{bmatrix}$$

$$T^{-1}(x_1, x_2) = \left(\frac{1}{3}x_1 + \frac{1}{12}x_2, \frac{1}{4}x_2 \right)$$

Example

Suppose $T : \mathbb{R}^n \longrightarrow \mathbb{R}^n$ is a one to one linear transformation. Can we determine whether T is onto? Why (or why not)?

Yes, we can say T is onto. Its standard matrix is $n \times n$. If $x \mapsto Ax$ is one to one, it must be onto.