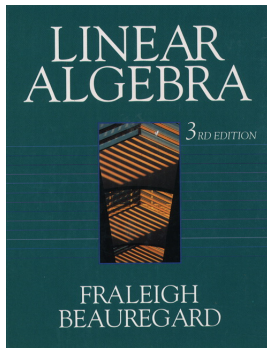


# Linear Algebra

## Chapter 1. Vectors, Matrices, and Linear Systems

### Section 1.5. Inverses of Square Matrices—Proofs of Theorems



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## Example 1.5.A

**Example 1.5.A.** It is easy to invert an elementary matrix. For example, suppose  $E_1$  interchanges Row 1 and Row 2 of a  $3 \times 3$  matrix. Suppose  $E_2$  multiplies Row 2 by 7 in a  $3 \times 3$  matrix. Find the inverses of  $E_1$  and  $E_2$ .

**Solution.** We have  $E_1 = \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}$  and  $E_2 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 7 & 0 \\ 0 & 0 & 1 \end{bmatrix}$ .

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Let  $A$  be an  $n \times n$  matrix. The linear system  $A\vec{x} = \vec{b}$  has a solution for every choice of column vector  $\vec{b} \in \mathbb{R}^n$  if and only if  $A$  is row equivalent to the  $n \times n$  identity matrix  $\mathcal{I}$ .

**Proof.** Suppose  $A$  is row equivalent to  $\mathcal{I}$ . Let  $\vec{b}$  be any column vector in  $\mathbb{R}^n$ . Then  $[A \mid \vec{b}] \sim [\mathcal{I} \mid \vec{c}]$  for some column vector  $\vec{c} \in \mathbb{R}^n$ . Then, by Theorem 1.6,  $\vec{x} = \vec{c}$  is a solution to  $A\vec{x} = \vec{b}$ .

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Suppose  $A$  is not row equivalent to  $\mathcal{I}$ . Row reduce  $A$  to a reduced row echelon form  $H$  (so  $H \neq \mathcal{I}$ ). So the last row (i.e., the  $n$ th row) of  $H$  must be all zeros.



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$$\begin{aligned} [E_t \cdots E_2 E_1 A \mid E_t \cdots E_2 E_1 \vec{b}] &= [E_t \cdots E_2 E_1 A \mid (E_t \cdots E_2 E_1)(E_t \cdots E_2 E_1)^{-1} \vec{e}_n] \\ &= [H \mid \vec{e}_n]. \end{aligned}$$

But then the last row of  $H$  consists of all zeros to the left of the partition and 1 to the right of the partition. So by Theorem 1.7(1), “Solutions of  $A\vec{x} = \vec{b}$ ,”  $A\vec{x} = \vec{b}$  has no solution. So if  $A$  is not row equivalent to  $\mathcal{I}$  then the system  $A\vec{x} = \vec{b}$  does not have a solution for all  $\vec{b} \in \mathbb{R}^n$ .  $\square$

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**Page 84 Number 12.** Determine whether the span of the column vectors

$$\text{of } A = \begin{bmatrix} 1 & -2 & 1 & 0 \\ -3 & 5 & 0 & 2 \\ 0 & 1 & 2 & -4 \\ -1 & 2 & 4 & -2 \end{bmatrix} \text{ span } \mathbb{R}^4.$$

**Solution.** Recall that for any  $\vec{x} \in \mathbb{R}^n$ ,  $A\vec{x}$  is a linear combination of the columns of  $A$  by Note 1.3.A. So to see if the column vectors of  $A$  span  $\mathbb{R}^4$ , we need to choose an arbitrary  $\vec{b} \in \mathbb{R}^4$  and see if there is  $\vec{x} \in \mathbb{R}^4$  such that  $A\vec{x} = \vec{b}$ . That is, we need to see if  $A\vec{x} = \vec{b}$  has a solution for every  $\vec{b} \in \mathbb{R}^4$ .

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Solution (continued).

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Now  $H$  is in reduced row echelon form and  $H \neq \mathcal{I}$ . So Lemma 1.1 implies that NO, the columns do not span  $\mathbb{R}^4$ .  $\square$

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# Theorem 1.11

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Let  $A$  and  $C$  be  $n \times n$  matrices. Then  $CA = \mathcal{I}$  if and only if  $AC = \mathcal{I}$ .

**Proof.** Suppose that  $AC = \mathcal{I}$ . Then the equation  $A\vec{x} = \vec{b}$  has a solution for every column vector  $\vec{b} \in \mathbb{R}^n$ . Notice that  $\vec{x} = C\vec{b}$  is a solution because

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solve the system 
$$\begin{aligned} 6x_1 + 7x_2 &= 4 \\ 8x_1 + 9x_2 &= 6. \end{aligned}$$

**Solution.** We form  $[A|\mathcal{I}]$  and apply Gauss-Jordan elimination to produce the row equivalent  $[\mathcal{I}|A^{-1}]$  (if possible).



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$$\left[ \begin{array}{cc|cc} 6 & 7 & 1 & 0 \\ 8 & 9 & 0 & 1 \end{array} \right] \xrightarrow{R_1 \rightarrow R_1/6} \left[ \begin{array}{cc|cc} 1 & 7/6 & 1/6 & 0 \\ 8 & 9 & 0 & 1 \end{array} \right]$$

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## Page 84 Number 4 (continued 1)

**Solution (continued).**

$$\left[ \begin{array}{cc|cc} 1 & 7/6 & 1/6 & 0 \\ 0 & 1 & 4 & -3 \end{array} \right] \xrightarrow{R_1 \rightarrow R_1 - (7/6)R_2}$$

$$\left[ \begin{array}{cc|cc} 1 - (7/6)(0) & 7/6 - (7/6)(1) & 1/6 - (7/6)(4) & 0 - (7/6)(-3) \\ 0 & 1 & 4 & -3 \end{array} \right]$$

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So  $A^{-1} = \begin{bmatrix} -9/2 & 7/2 \\ 4 & -3 \end{bmatrix}.$

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So  $A^{-1} = \boxed{\begin{bmatrix} -9/2 & 7/2 \\ 4 & -3 \end{bmatrix}}.$

## Page 84 Number 4 (continued 2)

**Solution (continued).** For the system of equations, we express it as a matrix product  $A\vec{x} = \vec{b}$ :  $\begin{bmatrix} 6 & 7 \\ 8 & 9 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 4 \\ 6 \end{bmatrix}$ . Then  $A^{-1}A\vec{x} = A^{-1}\vec{b}$  or  $I\vec{x} = A^{-1}\vec{b}$  or  $\vec{x} = A^{-1}\vec{b}$ . So

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and the solution is  $x_1 = 3, x_2 = -2$ .  $\square$

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**Solution.** We know that  $(AB)^T = B^T A^T$  (see “Properties of the Transpose Operator” in Section 1.3; page 4 of the notes).

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**Page 86 number 30.** A square matrix  $A$  is said to be *idempotent* if  $A^2 = A$ .

**(a)** Give an example of an idempotent matrix other than  $0$  and  $\mathcal{I}$ .

**Solution.** An easy example can be found by slightly modifying  $\mathcal{I}$ .

Consider, say,  $A = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}$ . Then

$$A^2 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix} = A.$$

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