

MAP 2210 – Aplicações de Álgebra Linear

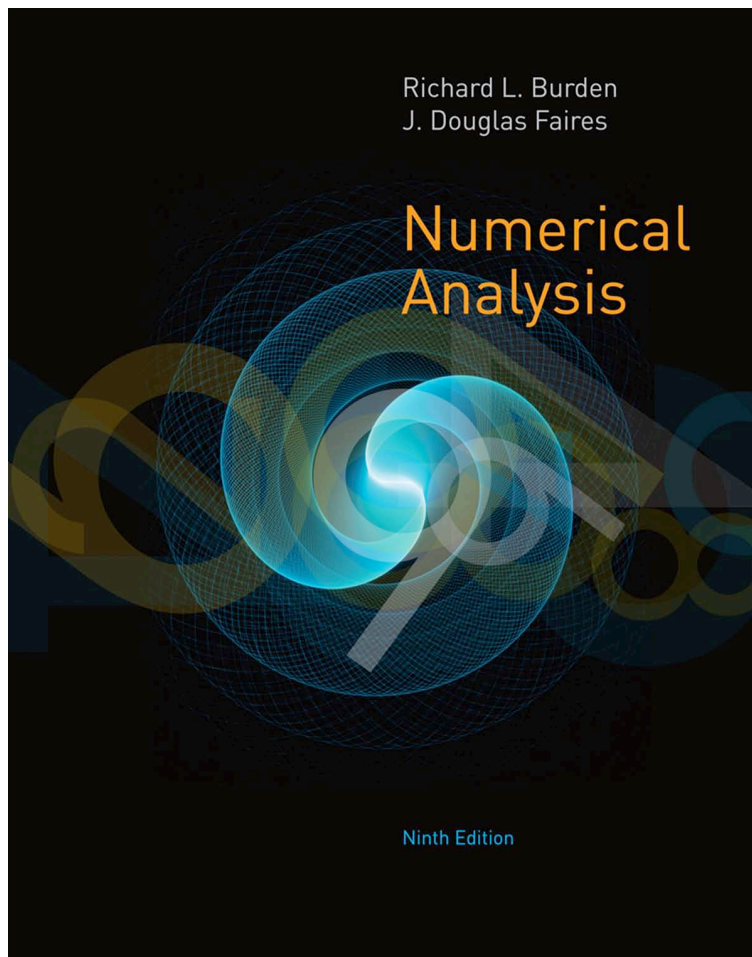
1º Semestre - 2020

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Objetivos

Formação básica de álgebra linear aplicada a problemas numéricos.
Resolução de problemas em microcomputadores usando linguagens e/ou software adequados fora do horário de aula.



Numerical Analysis

NINTH EDITION

Richard L. Burden

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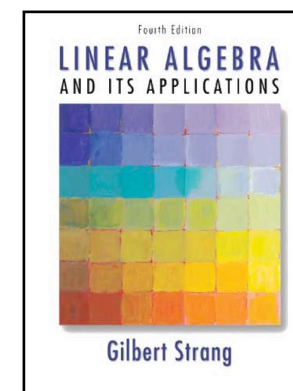
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6.3 Linear Algebra and Matrix Inversion

- Definition 6.1** An $n \times m$ (n by m) **matrix** is a rectangular array of elements with n rows and m columns in which not only is the value of an element important, but also its position in the array. ■
- Definition 6.2** Two matrices A and B are **equal** if they have the same number of rows and columns, say $n \times m$, and if $a_{ij} = b_{ij}$, for each $i = 1, 2, \dots, n$ and $j = 1, 2, \dots, m$. ■
- Definition 6.3** If A and B are both $n \times m$ matrices, then the **sum** of A and B , denoted $A + B$, is the $n \times m$ matrix whose entries are $a_{ij} + b_{ij}$, for each $i = 1, 2, \dots, n$ and $j = 1, 2, \dots, m$. ■
- Definition 6.4** If A is an $n \times m$ matrix and λ is a real number, then the **scalar multiplication** of λ and A , denoted λA , is the $n \times m$ matrix whose entries are λa_{ij} , for each $i = 1, 2, \dots, n$ and $j = 1, 2, \dots, m$. ■

We have the following general properties for matrix addition and scalar multiplication. These properties are sufficient to classify the set of all $n \times m$ matrices with real entries as a **vector space** over the field of real numbers.

- We let O denote a matrix all of whose entries are 0 and $-A$ denote the matrix whose entries are $-a_{ij}$.

Theorem 6.5 Let A , B , and C be $n \times m$ matrices and λ and μ be real numbers. The following properties of addition and scalar multiplication hold:

- | | |
|--|---|
| (i) $A + B = B + A$, | (ii) $(A + B) + C = A + (B + C)$, |
| (iii) $A + O = O + A = A$, | (iv) $A + (-A) = -A + A = O$, |
| (v) $\lambda(A + B) = \lambda A + \lambda B$, | (vi) $(\lambda + \mu)A = \lambda A + \mu A$, |
| (vii) $\lambda(\mu A) = (\lambda\mu)A$, | (viii) $1A = A$. |

All these properties follow from similar results concerning the real numbers. ■

Matrix-Vector Products

The product of matrices can also be defined in certain instances. We will first consider the product of an $n \times m$ matrix and a $m \times 1$ column vector.

Definition 6.6 Let A be an $n \times m$ matrix and \mathbf{b} an m -dimensional column vector. The **matrix-vector product** of A and \mathbf{b} , denoted $A\mathbf{b}$, is an n -dimensional column vector given by

$$A\mathbf{b} = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1m} \\ a_{21} & a_{22} & \cdots & a_{2m} \\ \vdots & \vdots & & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nm} \end{bmatrix} \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_m \end{bmatrix} = \begin{bmatrix} \sum_{i=1}^m a_{1i}b_i \\ \sum_{i=1}^m a_{2i}b_i \\ \vdots \\ \sum_{i=1}^m a_{ni}b_i \end{bmatrix}.$$

For this product to be defined the number of columns of the matrix A must match the number of rows of the vector \mathbf{b} , and the result is another column vector with the number of rows matching the number of rows in the matrix.

Matrix-Matrix Products

We can use this matrix-vector multiplication to define general matrix-matrix multiplication.

Definition 6.7 Let A be an $n \times m$ matrix and B an $m \times p$ matrix. The **matrix product** of A and B , denoted AB , is an $n \times p$ matrix C whose entries c_{ij} are

$$c_{ij} = \sum_{k=1}^m a_{ik}b_{kj} = a_{i1}b_{1j} + a_{i2}b_{2j} + \cdots + a_{im}b_{mj},$$

for each $i = 1, 2, \dots, n$, and $j = 1, 2, \dots, p$. ■

The computation of c_{ij} can be viewed as the multiplication of the entries of the i th row of A with corresponding entries in the j th column of B , followed by a summation; that is,

$$[a_{i1}, a_{i2}, \dots, a_{im}] \begin{bmatrix} b_{1j} \\ b_{2j} \\ \vdots \\ b_{mj} \end{bmatrix} = c_{ij},$$

where

$$c_{ij} = a_{i1}b_{1j} + a_{i2}b_{2j} + \cdots + a_{im}b_{mj} = \sum_{k=1}^m a_{ik}b_{kj}.$$

This explains why the number of columns of A must equal the number of rows of B for the product AB to be defined.

O produto de matrizes é comutativo ?

Definition 6.9

- (i) A **square** matrix has the same number of rows as columns.
- (ii) A **diagonal** matrix $D = [d_{ij}]$ is a square matrix with $d_{ij} = 0$ whenever $i \neq j$.
- (iii) The **identity matrix of order n** , $I_n = [\delta_{ij}]$, is a diagonal matrix whose diagonal entries are all 1s. When the size of I_n is clear, this matrix is generally written simply as I . ■

Definition 6.10

An **upper-triangular** $n \times n$ matrix $U = [u_{ij}]$ has, for each $j = 1, 2, \dots, n$, the entries

$$u_{ij} = 0, \quad \text{for each } i = j + 1, j + 2, \dots, n;$$

and a **lower-triangular** matrix $L = [l_{ij}]$ has, for each $j = 1, 2, \dots, n$, the entries

$$l_{ij} = 0, \quad \text{for each } i = 1, 2, \dots, j - 1. \quad \blacksquare$$

Inverse Matrices

Definition 6.11

An $n \times n$ matrix A is said to be **nonsingular** (or *invertible*) if an $n \times n$ matrix A^{-1} exists with $AA^{-1} = A^{-1}A = I$. The matrix A^{-1} is called the **inverse** of A . A matrix without an inverse is called **singular** (or *noninvertible*). ■

Theorem 6.12 For any nonsingular $n \times n$ matrix A :

- (i) A^{-1} is unique.
- (ii) A^{-1} is nonsingular and $(A^{-1})^{-1} = A$.
- (iii) If B is also a nonsingular $n \times n$ matrix, then $(AB)^{-1} = B^{-1}A^{-1}$. ■

Como encontrar a inversa ?

To find a method of computing A^{-1} assuming A is nonsingular, let us look again at matrix multiplication. Let B_j be the j th column of the $n \times n$ matrix B ,

$$B_j = \begin{bmatrix} b_{1j} \\ b_{2j} \\ \vdots \\ b_{nj} \end{bmatrix}.$$

If $AB = C$, then the j th column of C is given by the product

$$\begin{bmatrix} c_{1j} \\ c_{2j} \\ \vdots \\ c_{nj} \end{bmatrix} = C_j = AB_j = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{bmatrix} \begin{bmatrix} b_{1j} \\ b_{2j} \\ \vdots \\ b_{nj} \end{bmatrix} = \begin{bmatrix} \sum_{k=1}^n a_{1k} b_{kj} \\ \sum_{k=1}^n a_{2k} b_{kj} \\ \vdots \\ \sum_{k=1}^n a_{nk} b_{kj} \end{bmatrix}.$$

Suppose that A^{-1} exists and that $A^{-1} = B = (b_{ij})$. Then $AB = I$ and

$$AB_j = \begin{bmatrix} 0 \\ \vdots \\ 0 \\ 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}, \quad \text{where the value 1 appears in the } j\text{th row.}$$

To find B we need to solve n linear systems in which the j th column of the inverse is the solution of the linear system with right-hand side the j th column of I . The next illustration demonstrates this method.

Illustration

determine the inverse of the matrix

$$A = \begin{bmatrix} 1 & 2 & -1 \\ 2 & 1 & 0 \\ -1 & 1 & 2 \end{bmatrix},$$



$$A^{-1} = \begin{bmatrix} -\frac{2}{9} & \frac{5}{9} & -\frac{1}{9} \\ \frac{4}{9} & -\frac{1}{9} & \frac{2}{9} \\ -\frac{1}{3} & \frac{1}{3} & \frac{1}{3} \end{bmatrix}$$

As we saw in the illustration, in order to compute A^{-1} it is convenient to set up a larger augmented matrix,

$$\left[\begin{array}{ccc|ccc} A & \vdots & I \end{array} \right].$$

Upon performing the elimination in accordance with Algorithm 6.1, we obtain an augmented matrix of the form

$$\left[\begin{array}{ccc|ccc} U & \vdots & Y \end{array} \right],$$

where U is an upper-triangular matrix and Y is the matrix obtained by performing the same operations on the identity I that were performed to take A into U .

Gaussian elimination with backward substitution requires

$$\frac{4}{3}n^3 - \frac{1}{3}n \text{ multiplications/divisions} \quad \text{and} \quad \frac{4}{3}n^3 - \frac{3}{2}n^2 + \frac{n}{6} \text{ additions/subtractions.}$$

to solve the n linear systems (see Exercise 8(a)).

Transpose of a Matrix

Definition 6.13

The **transpose** of an $n \times m$ matrix $A = [a_{ij}]$ is the $m \times n$ matrix $A^t = [a_{ji}]$, where for each i , the i th column of A^t is the same as the i th row of A . A square matrix A is called **symmetric** if $A = A^t$. ■

Theorem 6.14

The following operations involving the transpose of a matrix hold whenever the operation is possible:

- | | |
|--------------------------------|---|
| (i) $(A^t)^t = A$, | (iii) $(AB)^t = B^t A^t$, |
| (ii) $(A + B)^t = A^t + B^t$, | (iv) if A^{-1} exists, then $(A^{-1})^t = (A^t)^{-1}$. ■ |

6.4 The Determinant of a Matrix

The *determinant* of a matrix provides existence and uniqueness results for linear systems having the same number of equations and unknowns. We will denote the determinant of a square matrix A by $\det A$, but it is also common to use the notation $|A|$.

Definition 6.15

Suppose that A is a square matrix.

- (i) If $A = [a]$ is a 1×1 matrix, then $\det A = a$.
- (ii) If A is an $n \times n$ matrix, with $n > 1$ the **minor** M_{ij} is the determinant of the $(n-1) \times (n-1)$ submatrix of A obtained by deleting the i th row and j th column of the matrix A .
- (iii) The **cofactor** A_{ij} associated with M_{ij} is defined by $A_{ij} = (-1)^{i+j} M_{ij}$.
- (iv) The **determinant** of the $n \times n$ matrix A , when $n > 1$, is given either by

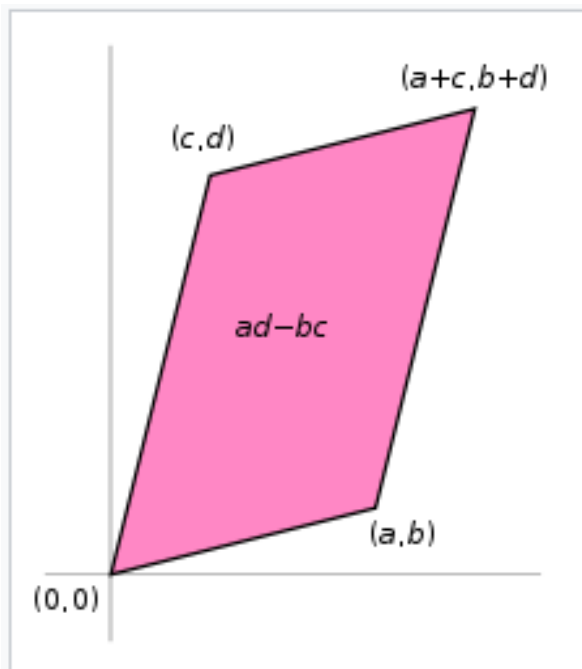
$$\det A = \sum_{j=1}^n a_{ij} A_{ij} = \sum_{j=1}^n (-1)^{i+j} a_{ij} M_{ij}, \quad \text{for any } i = 1, 2, \dots, n,$$

or by

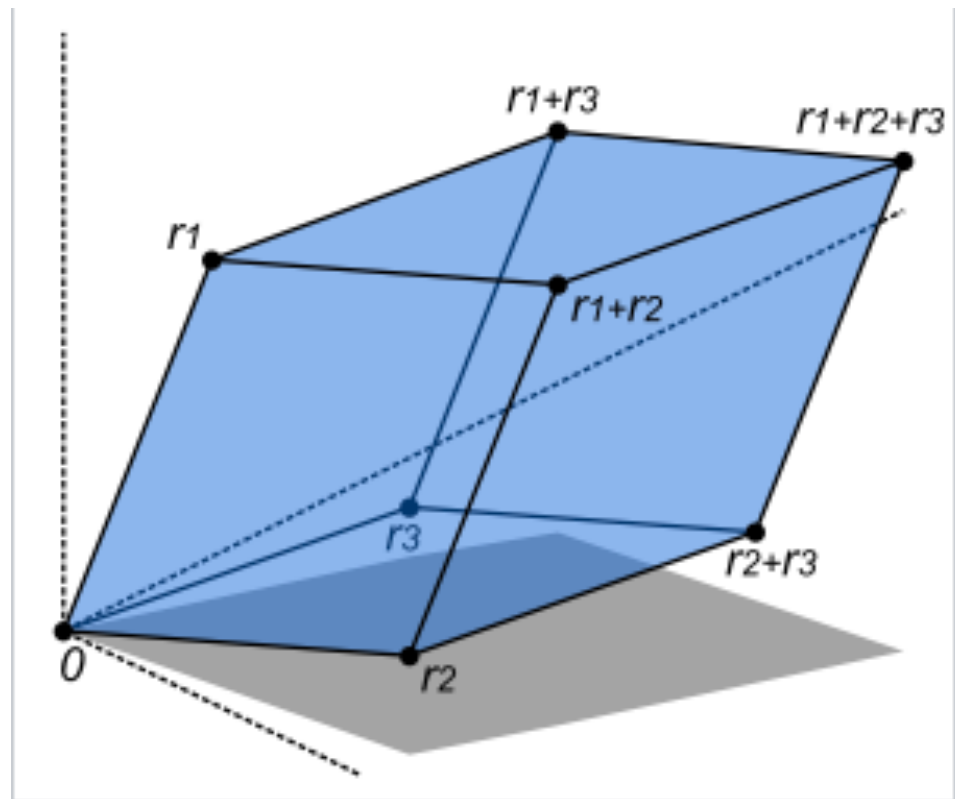
$$\det A = \sum_{i=1}^n a_{ij} A_{ij} = \sum_{i=1}^n (-1)^{i+j} a_{ij} M_{ij}, \quad \text{for any } j = 1, 2, \dots, n. \quad \blacksquare$$

$$|A| = \begin{vmatrix} a & b \\ c & d \end{vmatrix} = ad - bc.$$

$$\begin{aligned} |A| &= \begin{vmatrix} a & b & c \\ d & e & f \\ g & h & i \end{vmatrix} = a \begin{vmatrix} \square & \square & \square \\ \square & e & f \\ \square & h & i \end{vmatrix} - b \begin{vmatrix} \square & \square & \square \\ d & \square & f \\ g & \square & i \end{vmatrix} + c \begin{vmatrix} \square & \square & \square \\ d & e & \square \\ g & h & \square \end{vmatrix} \\ &= a \begin{vmatrix} e & f \\ h & i \end{vmatrix} - b \begin{vmatrix} d & f \\ g & i \end{vmatrix} + c \begin{vmatrix} d & e \\ g & h \end{vmatrix} \\ &= aei + bfg + cdh - ceg - bdi - afh. \end{aligned}$$



$$|A| = \begin{vmatrix} a & b \\ c & d \end{vmatrix} = ad - bc.$$



$$\begin{aligned} |A| &= \begin{vmatrix} a & b & c \\ d & e & f \\ g & h & i \end{vmatrix} = a \begin{vmatrix} \square & \square & \square \\ \square & e & f \\ \square & h & i \end{vmatrix} - b \begin{vmatrix} \square & \square & \square \\ d & \square & f \\ g & \square & i \end{vmatrix} + c \begin{vmatrix} \square & \square & \square \\ d & e & \square \\ g & h & \square \end{vmatrix} \\ &= a \begin{vmatrix} e & f \\ h & i \end{vmatrix} - b \begin{vmatrix} d & f \\ g & i \end{vmatrix} + c \begin{vmatrix} d & e \\ g & h \end{vmatrix} \\ &= aei + bfg + cdh - ceg - bdi - afh. \end{aligned}$$

Example 2

Compute the determinant of the matrix

$$A = \begin{bmatrix} 2 & 1 & -1 & 1 \\ 1 & 1 & 0 & 3 \\ -1 & 2 & 3 & -1 \\ 3 & -1 & -1 & 2 \end{bmatrix}$$

Grupo A: Use a linha 2

Grupo B: Use a coluna 3



Suppose A is an $n \times n$ matrix:

- (i) If any row or column of A has only zero entries, then $\det A = 0$.
- (ii) If A has two rows or two columns the same, then $\det A = 0$.
- (iii) If \tilde{A} is obtained from A by the operation $(E_i) \leftrightarrow (E_j)$, with $i \neq j$, then $\det \tilde{A} = -\det A$.
- (iv) If \tilde{A} is obtained from A by the operation $(\lambda E_i) \rightarrow (E_i)$, then $\det \tilde{A} = \lambda \det A$.
- (v) If \tilde{A} is obtained from A by the operation $(E_i + \lambda E_j) \rightarrow (E_i)$ with $i \neq j$, then $\det \tilde{A} = \det A$.
- (vi) If B is also an $n \times n$ matrix, then $\det AB = \det A \det B$.
- (vii) $\det A^t = \det A$.
- (viii) When A^{-1} exists, $\det A^{-1} = (\det A)^{-1}$.
- (ix) If A is an upper triangular, lower triangular, or diagonal matrix, then $\det A = \prod_{i=1}^n a_{ii}$. ■

It can be shown (see Exercise 9) that to calculate the determinant of a general $n \times n$ matrix by this definition requires $O(n!)$ multiplications/divisions and additions/subtractions.

Example 2

Compute the determinant of the matrix

$$A = \begin{bmatrix} 2 & 1 & -1 & 1 \\ 1 & 1 & 0 & 3 \\ -1 & 2 & 3 & -1 \\ 3 & -1 & -1 & 2 \end{bmatrix}$$

Usando as regras do Teorema 6.16

Fourth Edition

LINEAR ALGEBRA AND ITS APPLICATIONS

Gilbert Strang

4

Determinants

4.4 APPLICATIONS OF DETERMINANTS

This section follows through on four major applications: *inverse of A* , *solving $Ax = b$* , *volumes of boxes*, and *pivots*. They are among the key computations in linear algebra (done by elimination). Determinants give formulas for the answers.

1. **Computation of A^{-1} .** The 2 by 2 case shows how cofactors go into A^{-1} :

$$\begin{bmatrix} a & b \\ c & d \end{bmatrix}^{-1} = \frac{1}{ad - bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix} = \frac{1}{\det A} \begin{bmatrix} C_{11} & C_{21} \\ C_{12} & C_{22} \end{bmatrix}.$$

We are dividing by the determinant, and A is invertible exactly when $\det A$ is nonzero. The number $C_{11} = d$ is the cofactor of a . The number $C_{12} = -c$ is the cofactor of b (note the minus sign). That number C_{12} goes in row 2, column 1!

The row a, b times the column C_{11}, C_{12} produces $ad - bc$. This is the cofactor expansion of $\det A$. That is the clue we need: A^{-1} **divides the cofactors by $\det A$** .

**Cofactor matrix
 C is transposed**

$$A^{-1} = \frac{C^T}{\det A} \quad \text{means} \quad (A^{-1})_{ij} = \frac{C_{ji}}{\det A}. \quad (1)$$

Our goal is to verify this formula for A^{-1} . We have to see why $AC^T = (\det A)I$:

$$\begin{bmatrix} a_{11} & \cdots & a_{1n} \\ \vdots & & \vdots \\ a_{n1} & \cdots & a_{nn} \end{bmatrix} \begin{bmatrix} C_{11} & \cdots & C_{n1} \\ \vdots & & \vdots \\ C_{1n} & \cdots & C_{nn} \end{bmatrix} = \begin{bmatrix} \det A & \cdots & 0 \\ \vdots & & \vdots \\ 0 & \cdots & \det A \end{bmatrix}. \quad (2)$$

With cofactors C_{11}, \dots, C_{1n} in the first *column* and not the first row, they multiply a_{11}, \dots, a_{1n} and give the diagonal entry $\det A$. Every row of A multiplies its cofactors (*the cofactor expansion*) to give the same answer $\det A$ on the diagonal.

2. The Solution of $Ax = b$. The multiplication $x = A^{-1}b$ is just $C^T b$ divided by $\det A$. There is a famous way in which to write the answer (x_1, \dots, x_n) :

4C *Cramer's rule:* The j th component of $x = A^{-1}b$ is the ratio

$$x_j = \frac{\det B_j}{\det A}, \quad \text{where } B_j = \begin{bmatrix} a_{11} & a_{12} & \mathbf{b}_1 & a_{1n} \\ \vdots & \vdots & \vdots & \vdots \\ a_{n1} & a_{n2} & \mathbf{b}_n & a_{nn} \end{bmatrix} \text{ has } b \text{ in column } j. \quad (4)$$

Proof Expand $\det B_j$ in cofactors of its j th column (which is b). Since the cofactors ignore that column, $\det B_j$ is exactly the j th component in the product $C^T b$:

$$\det B_j = b_1 C_{1j} + b_2 C_{2j} + \dots + b_n C_{nj}.$$

Dividing this by $\det A$ gives x_j . Each component of x is a *ratio of two determinants*. That fact might have been recognized from Gaussian elimination, but it never was. ■

The key result relating nonsingularity, Gaussian elimination, linear systems, and determinants is that the following statements are equivalent.

Theorem 6.17

The following statements are equivalent for any $n \times n$ matrix A :

- (i) The equation $A\mathbf{x} = \mathbf{0}$ has the unique solution $\mathbf{x} = \mathbf{0}$.
- (ii) The system $A\mathbf{x} = \mathbf{b}$ has a unique solution for any n -dimensional column vector \mathbf{b} .
- (iii) The matrix A is nonsingular; that is, A^{-1} exists.
- (iv) $\det A \neq 0$.
- (v) Gaussian elimination with row interchanges can be performed on the system $A\mathbf{x} = \mathbf{b}$ for any n -dimensional column vector \mathbf{b} . ■

7. Given the two 4×4 linear systems having the same coefficient matrix:

$$\begin{array}{rcl} x_1 - x_2 + 2x_3 - x_4 & = & 6, \\ x_1 & - & x_3 + x_4 = 4, \\ 2x_1 + x_2 + 3x_3 - 4x_4 & = & -2, \\ -x_2 + x_3 - x_4 & = & 5; \end{array} \qquad \begin{array}{rcl} x_1 - x_2 + 2x_3 - x_4 & = & 1, \\ x_1 & - & x_3 + x_4 = 1, \\ 2x_1 + x_2 + 3x_3 - 4x_4 & = & 2, \\ -x_2 + x_3 - x_4 & = & -1. \end{array}$$

- a. Solve the linear systems by applying Gaussian elimination to the augmented matrix

$$\left[\begin{array}{cccc|cc} 1 & -1 & 2 & -1 & 6 & 1 \\ 1 & 0 & -1 & 1 & 4 & 1 \\ 2 & 1 & 3 & -4 & -2 & 2 \\ 0 & -1 & 1 & -1 & 5 & -1 \end{array} \right].$$

E calcule o determinante da matriz A

Fim...

ALLA 03