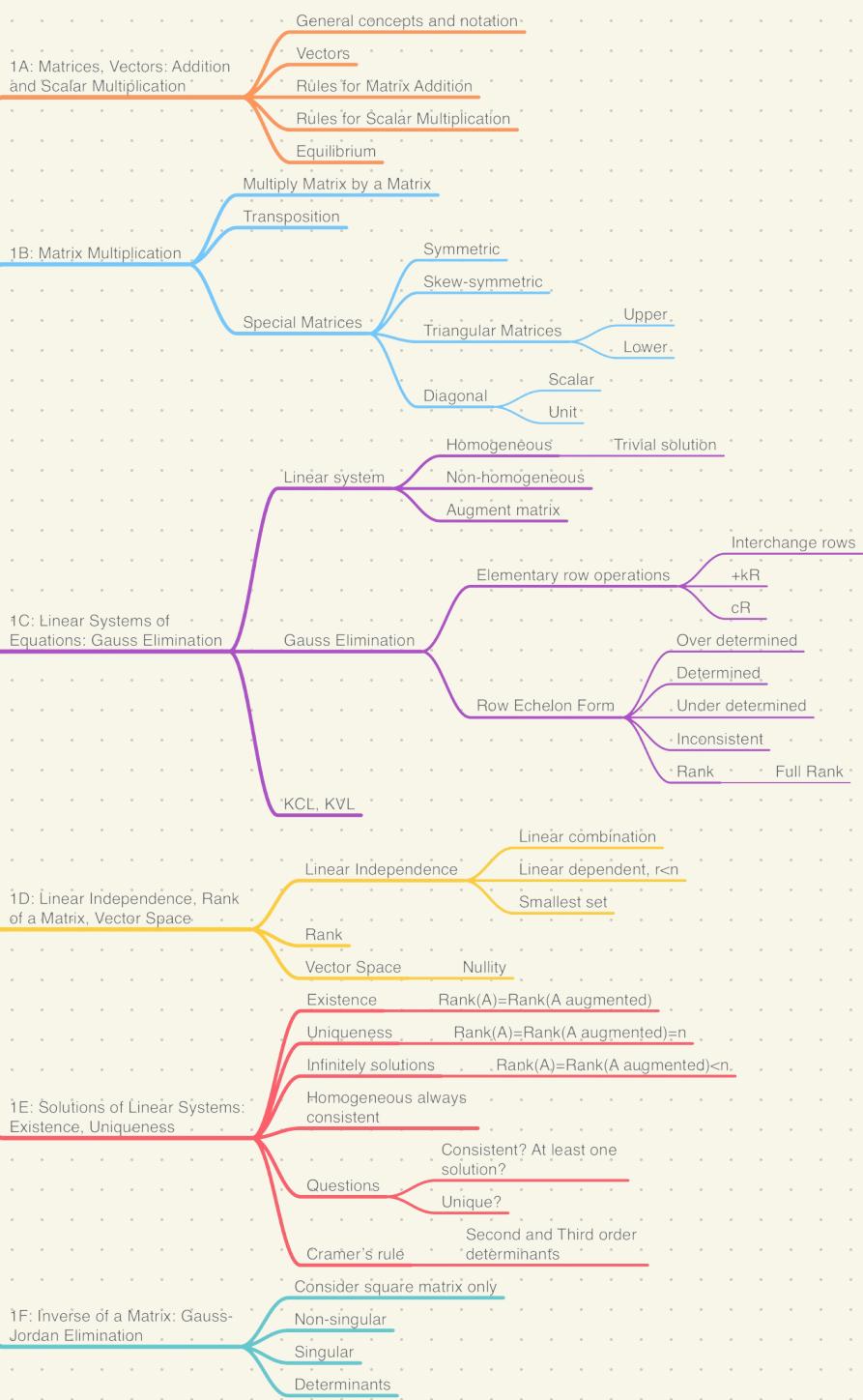


C7: Matrices and Linear Systems



Chapter 7

- Caution

Inverse of a Matrix by Determinants

The inverse of a nonsingular $n \times n$ matrix $\mathbf{A} = [a_{jk}]$ is given by

$$(4) \quad \mathbf{A}^{-1} = \frac{1}{\det \mathbf{A}} [\mathbf{C}_{jk}]^T = \frac{1}{\det \mathbf{A}} \begin{bmatrix} C_{11} & C_{21} & \cdots & C_{n1} \\ C_{12} & C_{22} & \cdots & C_{n2} \\ \vdots & \vdots & \ddots & \vdots \\ C_{1n} & C_{2n} & \cdots & C_{nn} \end{bmatrix},$$

where C_{jk} is the cofactor of a_{jk} in $\det \mathbf{A}$ (see Sec. 7.7). (CAUTION! Note well that in \mathbf{A}^{-1} , the cofactor C_{jk} occupies the same place as a_{kj} (not a_{jk}) does in \mathbf{A} .)

In particular, the inverse of

$$(4*) \quad \mathbf{A} = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \quad \text{is} \quad \mathbf{A}^{-1} = \frac{1}{\det \mathbf{A}} \begin{bmatrix} a_{22} & -a_{12} \\ -a_{21} & a_{11} \end{bmatrix}.$$

- (10)
- (a) $(\mathbf{A}^T)^T = \mathbf{A}$
 - (b) $(\mathbf{A} + \mathbf{B})^T = \mathbf{A}^T + \mathbf{B}^T$
 - (c) $(c\mathbf{A})^T = c\mathbf{A}^T$
 - (d) $(\mathbf{AB})^T = \mathbf{B}^T\mathbf{A}^T$.

CAUTION! Note that in (10d) the transposed matrices are *in reversed order*. We leave the proofs as an exercise in Probs. 9 and 10.

Elementary Row Operations for Matrices:

Interchange of two rows

Addition of a constant multiple of one row to another row

Multiplication of a row by a nonzero constant c

CAUTION! These operations are for rows, *not for columns*! They correspond to the following

CAUTION! $\det(c\mathbf{A}) = c^n \det \mathbf{A}$ (not $c \det \mathbf{A}$). Explain why.

$$(\mathbf{AC})^{-1} = \mathbf{C}^{-1}\mathbf{A}^{-1}.$$

THEOREM 1

Behavior of an n th-Order Determinant under Elementary Row Operations

- (a) *Interchange of two rows multiplies the value of the determinant by -1 .*
- (b) *Addition of a multiple of a row to another row does not alter the value of the determinant.*
- (c) *Multiplication of a row by a nonzero constant c multiplies the value of the determinant by c . (This holds also when $c = 0$, but no longer gives an elementary row operation.)*

-Theorems & Rules

Rules for Matrix Addition and Scalar Multiplication. From the familiar laws for the addition of numbers we obtain similar laws for the addition of matrices of the same size $m \times n$, namely,

- (3)
- (a) $\mathbf{A} + \mathbf{B} = \mathbf{B} + \mathbf{A}$
 - (b) $(\mathbf{A} + \mathbf{B}) + \mathbf{C} = \mathbf{A} + (\mathbf{B} + \mathbf{C})$ (written $\mathbf{A} + \mathbf{B} + \mathbf{C}$)
 - (c) $\mathbf{A} + \mathbf{0} = \mathbf{A}$
 - (d) $\mathbf{A} + (-\mathbf{A}) = \mathbf{0}.$

Hence matrix addition is *commutative* and *associative* [by (3a) and (3b)].

Similarly, for scalar multiplication we obtain the rules

- (4)
- (a) $c(\mathbf{A} + \mathbf{B}) = c\mathbf{A} + c\mathbf{B}$
 - (b) $(c + k)\mathbf{A} = c\mathbf{A} + k\mathbf{A}$
 - (c) $c(k\mathbf{A}) = (ck)\mathbf{A}$ (written $ck\mathbf{A}$)
 - (d) $1\mathbf{A} = \mathbf{A}.$

- (a) $(k\mathbf{A})\mathbf{B} = k(\mathbf{AB}) = \mathbf{A}(k\mathbf{B})$ written $k\mathbf{AB}$ or \mathbf{AkB}
- (b) $\mathbf{A}(\mathbf{BC}) = (\mathbf{AB})\mathbf{C}$ written \mathbf{ABC}
- (c) $(\mathbf{A} + \mathbf{B})\mathbf{C} = \mathbf{AC} + \mathbf{BC}$
- (d) $\mathbf{C}(\mathbf{A} + \mathbf{B}) = \mathbf{CA} + \mathbf{CB}$

THEOREM 3

Rank in Terms of Column Vectors

The rank r of a matrix \mathbf{A} equals the maximum number of linearly independent column vectors of \mathbf{A} .

Hence \mathbf{A} and its transpose \mathbf{A}^T have the same rank.

Rank in Terms of Determinants

Consider an $m \times n$ matrix $\mathbf{A} = [a_{jk}]$:

- (1) \mathbf{A} has rank $r \geq 1$ if and only if \mathbf{A} has an $r \times r$ submatrix with a nonzero determinant.
- (2) The determinant of any square submatrix with more than r rows, contained in \mathbf{A} (if such a matrix exists!) has a value equal to zero.

Furthermore, if $m = n$, we have:

- (3) An $n \times n$ square matrix \mathbf{A} has rank n if and only if

$$\det \mathbf{A} \neq 0.$$

Cramer's Theorem (Solution of Linear Systems by Determinants)

- (a) If a linear system of n equations in the same number of unknowns x_1, \dots, x_n

$$(6) \quad \begin{aligned} a_{11}x_1 + a_{12}x_2 + \cdots + a_{1n}x_n &= b_1 \\ a_{21}x_1 + a_{22}x_2 + \cdots + a_{2n}x_n &= b_2 \\ \vdots &\vdots \\ a_{n1}x_1 + a_{n2}x_2 + \cdots + a_{nn}x_n &= b_n \end{aligned}$$

has a nonzero coefficient determinant $D = \det \mathbf{A}$, the system has precisely one solution. This solution is given by the formulas

$$(7) \quad x_1 = \frac{D_1}{D}, \quad x_2 = \frac{D_2}{D}, \quad \cdots, \quad x_n = \frac{D_n}{D} \quad (\text{Cramer's rule})$$

where D_k is the determinant obtained from D by replacing in D the k th column by the column with the entries b_1, \dots, b_n .

- (b) Hence if the system (6) is homogeneous and $D \neq 0$, it has only the trivial solution $x_1 = 0, x_2 = 0, \dots, x_n = 0$. If $D = 0$, the homogeneous system also has nontrivial solutions.

Existence of the Inverse

The inverse \mathbf{A}^{-1} of an $n \times n$ matrix \mathbf{A} exists if and only if $\text{rank } \mathbf{A} = n$, thus (by Theorem 3, Sec. 7.7) if and only if $\det \mathbf{A} \neq 0$. Hence \mathbf{A} is nonsingular if $\text{rank } \mathbf{A} = n$, and is singular if $\text{rank } \mathbf{A} < n$.

Determinant of a Product of Matrices

For any $n \times n$ matrices \mathbf{A} and \mathbf{B} ,

$$(10) \quad \det(\mathbf{AB}) = \det(\mathbf{BA}) = \det \mathbf{A} \det \mathbf{B}.$$

This product is called the **inner product** or **dot product** of \mathbf{a} and \mathbf{b} . Other notations for it are (\mathbf{a}, \mathbf{b}) and $\mathbf{a} \cdot \mathbf{b}$. Thus

$$\mathbf{a}^\top \mathbf{b} = (\mathbf{a}, \mathbf{b}) = \mathbf{a} \cdot \mathbf{b} = [a_1 \cdots a_n] \begin{bmatrix} b_1 \\ \vdots \\ b_n \end{bmatrix} = \sum_{i=1}^n a_i b_i = a_1 b_1 + \cdots + a_n b_n.$$

$$(3) \quad |(\mathbf{a}, \mathbf{b})| \leq \|\mathbf{a}\| \|\mathbf{b}\| \quad (\text{Cauchy-Schwarz inequality}).$$

From this follows

$$(4) \quad \|\mathbf{a} + \mathbf{b}\| \leq \|\mathbf{a}\| + \|\mathbf{b}\| \quad (\text{Triangle inequality}).$$

A simple direct calculation gives

$$(5) \quad \|\mathbf{a} + \mathbf{b}\|^2 + \|\mathbf{a} - \mathbf{b}\|^2 = 2(\|\mathbf{a}\|^2 + \|\mathbf{b}\|^2) \quad (\text{Parallelogram equality}).$$

$$\|\mathbf{a}\| = \sqrt{(\mathbf{a}, \mathbf{a})} = \sqrt{\mathbf{a}^\top \mathbf{a}} = \sqrt{a_1^2 + \cdots + a_n^2}.$$

Linear Algebra: Matrices, Vectors, Determinants.

Linear Systems

An $m \times n$ **matrix** $\mathbf{A} = [a_{jk}]$ is a rectangular array of numbers or functions (“entries,” “elements”) arranged in m horizontal **rows** and n vertical **columns**. If $m = n$, the matrix is called **square**. A $1 \times n$ matrix is called a **row vector** and an $m \times 1$ matrix a **column vector** (Sec. 7.1).

The **sum** $\mathbf{A} + \mathbf{B}$ of matrices of the same **size** (i.e., both $m \times n$) is obtained by adding corresponding entries. The **product** of \mathbf{A} by a scalar c is obtained by multiplying each a_{jk} by c (Sec. 7.1).

The **product** $\mathbf{C} = \mathbf{AB}$ of an $m \times n$ matrix \mathbf{A} by an $r \times p$ matrix $\mathbf{B} = [b_{jk}]$ is defined only when $r = n$, and is the $m \times p$ matrix $\mathbf{C} = [c_{jk}]$ with entries

$$(1) \quad c_{jk} = a_{j1}b_{1k} + a_{j2}b_{2k} + \cdots + a_{jn}b_{nk} \quad (\text{row } j \text{ of } \mathbf{A} \text{ times column } k \text{ of } \mathbf{B}).$$

This multiplication is motivated by the composition of **linear transformations** (Secs. 7.2, 7.9). It is associative, but is *not commutative*: if \mathbf{AB} is defined, \mathbf{BA} may not be defined, but even if \mathbf{BA} is defined, $\mathbf{AB} \neq \mathbf{BA}$ in general. Also $\mathbf{AB} = \mathbf{0}$ may not imply $\mathbf{A} = \mathbf{0}$ or $\mathbf{B} = \mathbf{0}$ or $\mathbf{BA} = \mathbf{0}$ (Secs. 7.2, 7.8). Illustrations:

$$\begin{aligned} \begin{bmatrix} 1 & 1 \\ 2 & 2 \end{bmatrix} \begin{bmatrix} -1 & 1 \\ 1 & -1 \end{bmatrix} &= \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} \\ \begin{bmatrix} -1 & 1 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 2 & 2 \end{bmatrix} &= \begin{bmatrix} 1 & 1 \\ -1 & -1 \end{bmatrix} \\ [1 \ 2] \begin{bmatrix} 3 \\ 4 \end{bmatrix} &= [11], \quad \begin{bmatrix} 3 \\ 4 \end{bmatrix} [1 \ 2] = \begin{bmatrix} 3 & 6 \\ 4 & 8 \end{bmatrix}. \end{aligned}$$

The **transpose** \mathbf{A}^T of a matrix $\mathbf{A} = [a_{jk}]$ is $\mathbf{A}^T = [a_{kj}]$; rows become columns and conversely (Sec. 7.2). Here, \mathbf{A} need not be square. If it is and $\mathbf{A} = \mathbf{A}^T$, then \mathbf{A} is called **symmetric**; if $\mathbf{A} = -\mathbf{A}^T$, it is called **skew-symmetric**. For a product, $(\mathbf{AB})^T = \mathbf{B}^T \mathbf{A}^T$ (Sec. 7.2).

A main application of matrices concerns **linear systems of equations**

$$(2) \quad \mathbf{Ax} = \mathbf{b} \quad (\text{Sec. 7.3})$$

(m equations in n unknowns x_1, \dots, x_n ; \mathbf{A} and \mathbf{b} given). The most important method of solution is the **Gauss elimination** (Sec. 7.3), which reduces the system to “triangular” form by *elementary row operations*, which leave the set of solutions unchanged. (Numeric aspects and variants, such as *Doolittle’s* and *Cholesky’s methods*, are discussed in Secs. 20.1 and 20.2.)

Cramer's rule (Secs. 7.6, 7.7) represents the unknowns in a system (2) of n equations in n unknowns as quotients of determinants; for numeric work it is impractical. **Determinants** (Sec. 7.7) have decreased in importance, but will retain their place in eigenvalue problems, elementary geometry, etc.

The **inverse** \mathbf{A}^{-1} of a square matrix satisfies $\mathbf{AA}^{-1} = \mathbf{A}^{-1}\mathbf{A} = \mathbf{I}$. It exists if and only if $\det \mathbf{A} \neq 0$. It can be computed by the *Gauss–Jordan elimination* (Sec. 7.8).

The **rank** r of a matrix \mathbf{A} is the maximum number of linearly independent rows or columns of \mathbf{A} or, equivalently, the number of rows of the largest square submatrix of \mathbf{A} with nonzero determinant (Secs. 7.4, 7.7).

The system (2) has solutions if and only if $\text{rank } \mathbf{A} = \text{rank } [\mathbf{A} \quad \mathbf{b}]$, where $[\mathbf{A} \quad \mathbf{b}]$ is the **augmented matrix** (Fundamental Theorem, Sec. 7.5).

The **homogeneous system**

$$(3) \quad \mathbf{Ax} = \mathbf{0}$$

has solutions $\mathbf{x} \neq \mathbf{0}$ (“nontrivial solutions”) if and only if $\text{rank } \mathbf{A} < n$, in the case $m = n$ equivalently if and only if $\det \mathbf{A} = 0$ (Secs. 7.6, 7.7).

Vector spaces, inner product spaces, and linear transformations are discussed in Sec. 7.9. See also Sec. 7.4.