

SCHAUM'S OUTLINE OF
THEORY AND PROBLEMS

OF

MATRICES
SI (METRIC) EDITION

BY

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CONTENTS

	Page
Chapter 1 MATRICES	1
Matrices. Equal matrices. Sums of matrices. Products of matrices. Products by partitioning.	
<hr/>	
Chapter 2 SOME TYPES OF MATRICES	10
Triangular matrices. Scalar matrices. Diagonal matrices. The identity matrix. Inverse of a matrix. Transpose of a matrix. Symmetric matrices. Skew-symmetric matrices. Conjugate of a matrix. Hermitian matrices. Skew-Hermitian matrices. Direct sums.	
<hr/>	
Chapter 3 DETERMINANT OF A SQUARE MATRIX	20
Determinants of orders 2 and 3. Properties of determinants. Minors and cofactors. Algebraic complements.	
<hr/>	
Chapter 4 EVALUATION OF DETERMINANTS	32
Expansion along a row or column. The Laplace expansion. Expansion along the first row and column. Determinant of a product. Derivative of a determinant.	
<hr/>	
Chapter 5 EQUIVALENCE	39
Rank of a matrix. Non-singular and singular matrices. Elementary transformations. Inverse of an elementary transformation. Equivalent matrices. Row canonical form. Normal form. Elementary matrices. Canonical sets under equivalence. Rank of a product.	
<hr/>	
Chapter 6 THE ADJOINT OF A SQUARE MATRIX	49
The adjoint. The adjoint of a product. Minor of an adjoint.	
<hr/>	
Chapter 7 THE INVERSE OF A MATRIX	55
Inverse of a diagonal matrix. Inverse from the adjoint. Inverse from elementary matrices. Inverse by partitioning. Inverse of symmetric matrices. Right and left inverses of $m \times n$ matrices.	
<hr/>	
Chapter 8 FIELDS	64
Number fields. General fields. Sub-fields. Matrices over a field.	

CONTENTS

		Page
Chapter 9	LINEAR DEPENDENCE OF VECTORS AND FORMS Vectors. Linear dependence of vectors, linear forms, polynomials, and matrices.	67
<hr/>		
Chapter 10	LINEAR EQUATIONS System of non-homogeneous equations. Solution using matrices. Cramer's rule. Systems of homogeneous equations.	75
<hr/>		
Chapter 11	VECTOR SPACES Vector spaces. Sub-spaces. Basis and dimension. Sum space. Intersection space. Null space of a matrix. Sylvester's laws of nullity. Bases and coordinates.	85
<hr/>		
Chapter 12	LINEAR TRANSFORMATIONS Singular and non-singular transformations. Change of basis. Invariant space. Permutation matrix.	94
<hr/>		
Chapter 13	VECTORS OVER THE REAL FIELD Inner product. Length. Schwarz inequality. Triangle inequality. Orthogonal vectors and spaces. Orthonormal basis. Gram-Schmidt orthogonalization process. The Gramian. Orthogonal matrices. Orthogonal transformations. Vector product.	100
<hr/>		
Chapter 14	VECTORS OVER THE COMPLEX FIELD Complex numbers. Inner product. Length. Schwarz inequality. Triangle inequality. Orthogonal vectors and spaces. Orthonormal basis. Gram-Schmidt orthogonalization process. The Gramian. Unitary matrices. Unitary transformations.	110
<hr/>		
Chapter 15	CONGRUENCE Congruent matrices. Congruent symmetric matrices. Canonical forms of real symmetric, skew-symmetric, Hermitian, skew-Hermitian matrices under congruence.	115
<hr/>		
Chapter 16	BILINEAR FORMS Matrix form. Transformations. Canonical forms. Cogredient transformations. Contragredient transformations. Factorable forms.	125
<hr/>		
Chapter 17	QUADRATIC FORMS Matrix form. Transformations. Canonical forms. Lagrange reduction. Sylvester's law of inertia. Definite and semi-definite forms. Principal minors. Regular form. Kronecker's reduction. Factorable forms.	131

CONTENTS

	Page
<p>Chapter 18 HERMITIAN FORMS</p> <p style="padding-left: 2em;">Matrix form. Transformations. Canonical forms. Definite and semi-definite forms.</p>	146
<hr/>	
<p>Chapter 19 THE CHARACTERISTIC EQUATION OF A MATRIX.....</p> <p style="padding-left: 2em;">Characteristic equation and roots. Invariant vectors and spaces.</p>	149
<hr/>	
<p>Chapter 20 SIMILARITY</p> <p style="padding-left: 2em;">Similar matrices. Reduction to triangular form. Diagonal matrices.</p>	156
<hr/>	
<p>Chapter 21 SIMILARITY TO A DIAGONAL MATRIX.....</p> <p style="padding-left: 2em;">Real symmetric matrices. Orthogonal similarity. Pairs of real quadratic forms. Hermitian matrices. Unitary similarity. Normal matrices. Spectral decomposition. Field of values.</p>	163
<hr/>	
<p>Chapter 22 POLYNOMIALS OVER A FIELD.....</p> <p style="padding-left: 2em;">Sum, product, quotient of polynomials. Remainder theorem. Greatest common divisor. Least common multiple. Relatively prime polynomials. Unique factorization.</p>	172
<hr/>	
<p>Chapter 23 LAMBDA MATRICES</p> <p style="padding-left: 2em;">The λ-matrix or matrix polynomial. Sums, products, and quotients. Remainder theorem. Cayley-Hamilton theorem. Derivative of a matrix.</p>	179
<hr/>	
<p>Chapter 24 SMITH NORMAL FORM</p> <p style="padding-left: 2em;">Smith normal form. Invariant factors. Elementary divisors.</p>	188
<hr/>	
<p>Chapter 25 THE MINIMUM POLYNOMIAL OF A MATRIX.....</p> <p style="padding-left: 2em;">Similarity invariants. Minimum polynomial. Derogatory and non-derogatory matrices. Companion matrix.</p>	196
<hr/>	
<p>Chapter 26 CANONICAL FORMS UNDER SIMILARITY.....</p> <p style="padding-left: 2em;">Rational canonical form. A second canonical form. Hypercompanion matrix. Jacobson canonical form. Classical canonical form. A reduction to rational canonical form.</p>	203
<hr/>	
<p style="padding-left: 4em;">INDEX</p>	215
<p style="padding-left: 4em;">INDEX OF SYMBOLS.....</p>	219

Preface

Elementary matrix algebra has now become an integral part of the mathematical background necessary for such diverse fields as electrical engineering and education, chemistry and sociology, as well as for statistics and pure mathematics. This book, in presenting the more essential material, is designed primarily to serve as a useful supplement to current texts and as a handy reference book for those working in the several fields which require some knowledge of matrix theory. Moreover, the statements of theory and principle are sufficiently complete that the book could be used as a text by itself.

The material has been divided into twenty-six chapters, since the logical arrangement is thereby not disturbed while the usefulness as a reference book is increased. This also permits a separation of the treatment of real matrices, with which the majority of readers will be concerned, from that of matrices with complex elements. Each chapter contains a statement of pertinent definitions, principles, and theorems, fully illustrated by examples. These, in turn, are followed by a carefully selected set of solved problems and a considerable number of supplementary exercises.

The student new to matrix algebra soon finds that the solutions of numerical exercises are disarmingly simple. Difficulties are likely to arise from the constant round of definition, theorem, proof. The trouble here is essentially a matter of lack of mathematical maturity, and normally to be expected, since usually the student's previous work in mathematics has been concerned with the solution of numerical problems while precise statements of principles and proofs of theorems have in large part been deferred for later courses. The aim of the present book is to enable the reader, if he persists through the introductory paragraphs and solved problems in any chapter, to develop a reasonable degree of self-assurance about the material.

The solved problems, in addition to giving more variety to the examples illustrating the theorems, contain most of the proofs of any considerable length together with representative shorter proofs. The supplementary problems call both for the solution of numerical exercises and for proofs. Some of the latter require only proper modifications of proofs given earlier; more important, however, are the many theorems whose proofs require but a few lines. Some are of the type frequently misnamed "obvious" while others will be found to call for considerable ingenuity. None should be treated lightly, however, for it is due precisely to the abundance of such theorems that elementary matrix algebra becomes a natural first course for those seeking to attain a degree of mathematical maturity. While the large number of these problems in any chapter makes it impractical to solve all of them before moving to the next, special attention is directed to the supplementary problems of the first two chapters. A mastery of these will do much to give the reader confidence to stand on his own feet thereafter.

The author wishes to take this opportunity to express his gratitude to the staff of the Schaum Publishing Company for their splendid cooperation.

FRANK AYRES, JR.

Carlisle, Pa.
October, 1962

Chapter 1

Matrices

A RECTANGULAR ARRAY OF NUMBERS enclosed by a pair of brackets, such as

$$(a) \begin{bmatrix} 2 & 3 & 7 \\ 1 & -1 & 5 \end{bmatrix} \quad \text{and} \quad (b) \begin{bmatrix} 1 & 3 & 1 \\ 2 & 1 & 4 \\ 4 & 7 & 6 \end{bmatrix},$$

and subject to certain rules of operations given below is called a **matrix**. The matrix (a) could be considered as the **coefficient matrix** of the system of homogeneous linear equations $\begin{cases} 2x + 3y + 7z = 0 \\ x - y + 5z = 0 \end{cases}$ or as the **augmented matrix** of the system of non-homogeneous linear equations $\begin{cases} 2x + 3y = 7 \\ x - y = 5 \end{cases}$.

Later, we shall see how the matrix may be used to obtain solutions of these systems. The matrix (b) could be given a similar interpretation or we might consider its rows as simply the coordinates of the points (1, 3, 1), (2, 1, 4), and (4, 7, 6) in ordinary space. The matrix will be used later to settle such questions as whether or not the three points lie in the same plane with the origin or on the same line through the origin.

In the matrix

$$(1.1) \quad \begin{matrix} & \text{column} \\ & \downarrow \downarrow \\ \begin{bmatrix} a_{11} & a_{12} & a_{13} & \dots & a_{1n} \\ a_{21} & a_{22} & a_{23} & \dots & a_{2n} \\ \dots & \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots & \dots \\ a_{m1} & a_{m2} & a_{m3} & \dots & a_{mn} \end{bmatrix} & \begin{matrix} \leftarrow \\ \leftarrow \\ \leftarrow \\ \leftarrow \\ \leftarrow \end{matrix} & \text{row} \end{matrix}$$

the numbers or functions a_{ij} are called its **elements**. In the double subscript notation, the first subscript indicates the **row** and the second subscript indicates the **column** in which the element stands. Thus, all elements in the second row have 2 as first subscript and all the elements in the fifth column have 5 as second subscript. A matrix of m rows and n columns is said to be of order " m by n " or $m \times n$.

(In indicating a matrix pairs of parentheses, (), and double bars, $\| \quad \|$, are sometimes used. We shall use the double bracket notation throughout.)

At times the matrix (1.1) will be called "the $m \times n$ matrix $[a_{ij}]$ " or "the $m \times n$ matrix $A = [a_{ij}]$ ". When the order has been established, we shall write simply "the matrix A ".

SQUARE MATRICES. When $m = n$, (1.1) is square and will be called a **square matrix of order n** or an **n -square matrix**.

In a square matrix, the elements $a_{11}, a_{22}, \dots, a_{nn}$ are called its **diagonal elements**.

The sum of the diagonal elements of a square matrix A is called the **trace of A** .

EQUAL MATRICES. Two matrices $A = [a_{ij}]$ and $B = [b_{ij}]$ are said to be equal ($A=B$) if and only if they have the same order and each element of one is equal to the corresponding element of the other, that is, if and only if

$$a_{ij} = b_{ij}, \quad (i = 1, 2, \dots, m; j = 1, 2, \dots, n)$$

Thus, two matrices are equal if and only if one is a duplicate of the other.

ZERO MATRIX. A matrix, every element of which is zero, is called a **zero matrix**. When A is a zero matrix and there can be no confusion as to its order, we shall write $A = 0$ instead of the $m \times n$ array of zero elements.

SUMS OF MATRICES. If $A = [a_{ij}]$ and $B = [b_{ij}]$ are two $m \times n$ matrices, their sum (difference), $A \pm B$ is defined as the $m \times n$ matrix $C = [c_{ij}]$, where each element of C is the sum (difference) of the corresponding elements of A and B . Thus, $A \pm B = [a_{ij} \pm b_{ij}]$.

Example 1. If $A = \begin{bmatrix} 1 & 2 & 3 \\ 0 & 1 & 4 \end{bmatrix}$ and $B = \begin{bmatrix} 2 & 3 & 0 \\ -1 & 2 & 5 \end{bmatrix}$ then

$$A + B = \begin{bmatrix} 1+2 & 2+3 & 3+0 \\ 0+(-1) & 1+2 & 4+5 \end{bmatrix} = \begin{bmatrix} 3 & 5 & 3 \\ -1 & 3 & 9 \end{bmatrix}$$

and

$$A - B = \begin{bmatrix} 1-2 & 2-3 & 3-0 \\ 0-(-1) & 1-2 & 4-5 \end{bmatrix} = \begin{bmatrix} -1 & -1 & 3 \\ 1 & -1 & -1 \end{bmatrix}$$

Two matrices of the same order are said to be conformable for addition or subtraction. Two matrices of different orders cannot be added or subtracted. For example, the matrices (a) and (b) above are non-conformable for addition and subtraction.

The sum of k matrices A is a matrix of the same order as A and each of its elements is k times the corresponding element of A . We define: If k is any scalar (we call k a scalar to distinguish it from $[k]$ which is a 1×1 matrix) then by $kA = Ak$ is meant the matrix obtained from A by multiplying each of its elements by k .

Example 2. If $A = \begin{bmatrix} 1 & -2 \\ 2 & 3 \end{bmatrix}$, then

$$A + A + A = \begin{bmatrix} 1 & -2 \\ 2 & 3 \end{bmatrix} + \begin{bmatrix} 1 & -2 \\ 2 & 3 \end{bmatrix} + \begin{bmatrix} 1 & -2 \\ 2 & 3 \end{bmatrix} = \begin{bmatrix} 3 & -6 \\ 6 & 9 \end{bmatrix} = 3A = A \cdot 3$$

$$\text{and} \quad -5A = \begin{bmatrix} -5(1) & -5(-2) \\ -5(2) & -5(3) \end{bmatrix} = \begin{bmatrix} -5 & 10 \\ -10 & -15 \end{bmatrix}$$

In particular, by $-A$, called the negative of A , is meant the matrix obtained from A by multiplying each of its elements by -1 or by simply changing the sign of all of its elements. For every A , we have $A + (-A) = 0$, where 0 indicates the zero matrix of the same order as A .

Assuming that the matrices A, B, C are conformable for addition, we state:

- (a) $A + B = B + A$ (commutative law)
- (b) $A + (B + C) = (A + B) + C$ (associative law)
- (c) $k(A + B) = kA + kB = (A + B)k$, k a scalar
- (d) There exists a matrix D such that $A + D = B$.

These laws are a result of the laws of elementary algebra governing the addition of numbers and polynomials. They show, moreover,

1. Conformable matrices obey the same laws of addition as the elements of these matrices.

MULTIPLICATION. By the product AB in that order of the $1 \times m$ matrix $A = [a_{11} \ a_{12} \ a_{13} \ \dots \ a_{1m}]$ and

the $m \times 1$ matrix $B = \begin{bmatrix} b_{11} \\ b_{21} \\ b_{31} \\ \vdots \\ b_{m1} \end{bmatrix}$ is meant the 1×1 matrix $C = [a_{11} b_{11} + a_{12} b_{21} + \dots + a_{1m} b_{m1}]$.

$$\text{That is, } [a_{11} \ a_{12} \ \dots \ a_{1m}] \cdot \begin{bmatrix} b_{11} \\ b_{21} \\ \vdots \\ b_{m1} \end{bmatrix} = [a_{11} b_{11} + a_{12} b_{21} + \dots + a_{1m} b_{m1}] = \left[\sum_{k=1}^m a_{1k} b_{k1} \right].$$

Note that the operation is **row by column**; each element of the row is multiplied into the corresponding element of the column and then the products are summed.

Example 3 (a) $[2 \ 3 \ 4] \begin{bmatrix} 1 \\ -1 \\ 2 \end{bmatrix} = [2(1) + 3(-1) + 4(2)] = [7]$

(b) $[3 \ -1 \ 4] \begin{bmatrix} -2 \\ 6 \\ 3 \end{bmatrix} = [-6 - 6 + 12] = 0$

By the product AB in that order of the $m \times p$ matrix $A = [a_{ij}]$ and the $p \times n$ matrix $B = [b_{ij}]$ is meant the $m \times n$ matrix $C = [c_{ij}]$ where

$$c_{ij} = a_{i1} b_{1j} + a_{i2} b_{2j} + \dots + a_{ip} b_{pj} = \sum_{k=1}^p a_{ik} b_{kj}, \quad (i = 1, 2, \dots, m; j = 1, 2, \dots, n).$$

Think of A as consisting of m rows and B as consisting of n columns. In forming $C = AB$ each row of A is multiplied once and only once into each column of B . The element c_{ij} of C is then the product of the i th row of A and the j th column of B .

Example 4.

$$A \ B = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \\ a_{31} & a_{32} \end{bmatrix} \begin{bmatrix} b_{11} & b_{12} \\ b_{21} & b_{22} \end{bmatrix} = \begin{bmatrix} a_{11} b_{11} + a_{12} b_{21} & a_{11} b_{12} + a_{12} b_{22} \\ a_{21} b_{11} + a_{22} b_{21} & a_{21} b_{12} + a_{22} b_{22} \\ a_{31} b_{11} + a_{32} b_{21} & a_{31} b_{12} + a_{32} b_{22} \end{bmatrix}$$

3×2
 2×2
 3×2

The product AB is defined or A is conformable to B for multiplication only when the number of columns of A is equal to the number of rows of B . If A is conformable to B for multiplication (AB is defined), B is not necessarily conformable to A for multiplication (BA may or may not be defined).

See Problems 3-4.

Assuming that A, B, C are conformable for the indicated sums and products, we have

(e) $A(B + C) = AB + AC$ (first distributive law)

(f) $(A + B)C = AC + BC$ (second distributive law)

(g) $A(BC) = (AB)C$ (associative law)

However,

(h) $AB \neq BA$, generally,

(i) $AB = 0$ does not necessarily imply $A = 0$ or $B = 0$,

(j) $AB = AC$ does not necessarily imply $B = C$.

See Problems 3-6

PRODUCTS BY PARTITIONING. Let $A = [a_{ij}]$ be of order $m \times p$ and $B = [b_{ij}]$ be of order $p \times n$. In
 NOT TO BE COVERED

$$A = \left[\begin{array}{c|c|c} (m_1 \times p_1) & (m_1 \times p_2) & (m_1 \times p_3) \\ \hline (m_2 \times p_1) & (m_2 \times p_2) & (m_2 \times p_3) \end{array} \right], \quad B = \left[\begin{array}{c|c} (p_1 \times n_1) & (p_1 \times n_2) \\ \hline (p_2 \times n_1) & (p_2 \times n_2) \\ \hline (p_3 \times n_1) & (p_3 \times n_2) \end{array} \right]$$

$$\text{or} \quad A = \left[\begin{array}{c|c|c} A_{11} & A_{12} & A_{13} \\ \hline A_{21} & A_{22} & A_{23} \end{array} \right], \quad B = \left[\begin{array}{c|c} B_{11} & B_{12} \\ \hline B_{21} & B_{22} \\ \hline B_{31} & B_{32} \end{array} \right]$$

In any such partitioning, it is necessary that the columns of A and the rows of B be partitioned in exactly the same way; however m_1, m_2, n_1, n_2 may be any non-negative (including 0) integers such that $m_1 + m_2 = m$ and $n_1 + n_2 = n$. Then

$$AB = \left[\begin{array}{cc} A_{11}B_{11} + A_{12}B_{21} + A_{13}B_{31} & A_{11}B_{12} + A_{12}B_{22} + A_{13}B_{32} \\ A_{21}B_{11} + A_{22}B_{21} + A_{23}B_{31} & A_{21}B_{12} + A_{22}B_{22} + A_{23}B_{32} \end{array} \right] = \left[\begin{array}{cc} C_{11} & C_{12} \\ C_{21} & C_{22} \end{array} \right] = C$$

Example 5. Compute AB , given $A = \begin{bmatrix} 2 & 1 & 0 \\ 3 & 2 & 0 \\ 1 & 0 & 1 \end{bmatrix}$ and $B = \begin{bmatrix} 1 & 1 & 1 & 0 \\ 2 & 1 & 1 & 0 \\ 2 & 3 & 1 & 2 \end{bmatrix}$

Partitioning so that

$$A = \left[\begin{array}{cc|c} A_{11} & A_{12} & \\ \hline A_{21} & A_{22} & \\ \hline 1 & 0 & 1 \end{array} \right] \quad \text{and} \quad B = \left[\begin{array}{cc|cc} B_{11} & B_{12} & & \\ \hline B_{21} & B_{22} & & \\ \hline 1 & 1 & 1 & 0 \\ 2 & 1 & 1 & 0 \\ 2 & 3 & 1 & 2 \end{array} \right]$$

we have $AB = \begin{bmatrix} A_{11}B_{11} + A_{12}B_{21} & A_{11}B_{12} + A_{12}B_{22} \\ A_{21}B_{11} + A_{22}B_{21} & A_{21}B_{12} + A_{22}B_{22} \end{bmatrix}$

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

$$= \begin{bmatrix} \begin{bmatrix} 4 & 3 & 3 \\ 7 & 5 & 5 \end{bmatrix} + \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} & \begin{bmatrix} 0 \\ 0 \end{bmatrix} + \begin{bmatrix} 0 \\ 0 \end{bmatrix} \\ \begin{bmatrix} 1 & 1 & 1 \end{bmatrix} + \begin{bmatrix} 2 & 3 & 1 \end{bmatrix} & \begin{bmatrix} 0 \\ 0 \end{bmatrix} + \begin{bmatrix} 2 \end{bmatrix} \end{bmatrix} = \begin{bmatrix} \begin{bmatrix} 4 & 3 & 3 \\ 7 & 5 & 5 \end{bmatrix} & \begin{bmatrix} 0 \\ 0 \end{bmatrix} \\ \begin{bmatrix} 3 & 4 & 2 \end{bmatrix} & \begin{bmatrix} 2 \end{bmatrix} \end{bmatrix} = \begin{bmatrix} 4 & 3 & 3 & 0 \\ 7 & 5 & 5 & 0 \\ 3 & 4 & 2 & 2 \end{bmatrix}$$

See also Problem 9.

Let A, B, C, \dots be n -square matrices. Let A be partitioned into matrices of the indicated orders

$$\left[\begin{array}{c|c|c|c} (p_1 \times p_1) & (p_1 \times p_2) & \dots & (p_1 \times p_s) \\ \hline (p_2 \times p_1) & (p_2 \times p_2) & \dots & (p_2 \times p_s) \\ \hline \dots & \dots & \dots & \dots \\ \hline (p_s \times p_1) & (p_s \times p_2) & \dots & (p_s \times p_s) \end{array} \right] = \left[\begin{array}{cccc} A_{11} & A_{12} & \dots & A_{1s} \\ A_{21} & A_{22} & \dots & A_{2s} \\ \dots & \dots & \dots & \dots \\ A_{s1} & A_{s2} & \dots & A_{ss} \end{array} \right]$$

and let B, C, \dots be partitioned in exactly the same manner. Then sums, differences, and products may be formed using the matrices $A_{11}, A_{12}, \dots; B_{11}, B_{12}, \dots; C_{11}, C_{12}, \dots$

SOLVED PROBLEMS

$$1. (a) \begin{bmatrix} 1 & 2 & -1 & 0 \\ 4 & 0 & 2 & 1 \\ 2 & -5 & 1 & 2 \end{bmatrix} + \begin{bmatrix} 3 & -4 & 1 & 2 \\ 1 & 5 & 0 & 3 \\ 2 & -2 & 3 & -1 \end{bmatrix} = \begin{bmatrix} 1+3 & 2+(-4) & -1+1 & 0+2 \\ 4+1 & 0+5 & 2+0 & 1+3 \\ 2+2 & -5+(-2) & 1+3 & 2+(-1) \end{bmatrix} = \begin{bmatrix} 4 & -2 & 0 & 2 \\ 5 & 5 & 2 & 4 \\ 4 & -7 & 4 & 1 \end{bmatrix}$$

$$(b) \begin{bmatrix} 1 & 2 & -1 & 0 \\ 4 & 0 & 2 & 1 \\ 2 & -5 & 1 & 2 \end{bmatrix} - \begin{bmatrix} 3 & -4 & 1 & 2 \\ 1 & 5 & 0 & 3 \\ 2 & -2 & 3 & -1 \end{bmatrix} = \begin{bmatrix} 1-3 & 2+4 & -1-1 & 0-2 \\ 4-1 & 0-5 & 2-0 & 1-3 \\ 2-2 & -5+2 & 1-3 & 2+1 \end{bmatrix} = \begin{bmatrix} -2 & 6 & -2 & -2 \\ 3 & -5 & 2 & -2 \\ 0 & -3 & -2 & 3 \end{bmatrix}$$

$$(c) 3 \begin{bmatrix} 1 & 2 & -1 & 0 \\ 4 & 0 & 2 & 1 \\ 2 & -5 & 1 & 2 \end{bmatrix} = \begin{bmatrix} 3 & 6 & -3 & 0 \\ 12 & 0 & 6 & 3 \\ 6 & -15 & 3 & 6 \end{bmatrix}$$

$$(d) - \begin{bmatrix} 1 & 2 & -1 & 0 \\ 4 & 0 & 2 & 1 \\ 2 & -5 & 1 & 2 \end{bmatrix} = \begin{bmatrix} -1 & -2 & 1 & 0 \\ -4 & 0 & -2 & -1 \\ -2 & 5 & -1 & -2 \end{bmatrix}$$

2. If $A = \begin{bmatrix} 1 & 2 \\ 3 & 4 \\ 5 & 6 \end{bmatrix}$ and $B = \begin{bmatrix} -3 & -2 \\ 1 & -5 \\ 4 & 3 \end{bmatrix}$, find $D = \begin{bmatrix} p & q \\ r & s \\ t & u \end{bmatrix}$ such that $A + B - D = 0$.

$$\text{If } A + B - D = \begin{bmatrix} 1-3-p & 2-2-q \\ 3+1-r & 4-5-s \\ 5+4-t & 6+3-u \end{bmatrix} = \begin{bmatrix} -2-p & -q \\ 4-r & -1-s \\ 9-t & 9-u \end{bmatrix} = \begin{bmatrix} 0 & 0 \\ 0 & 0 \\ 0 & 0 \end{bmatrix}, \quad -2-p = 0 \text{ and } p = -2, \quad 4-r = 0$$

and $r = 4, \dots$. Then $D = \begin{bmatrix} -2 & 0 \\ 4 & -1 \\ 9 & 9 \end{bmatrix} = A + B$.

$$3. (a) \begin{bmatrix} 4 & 5 & 6 \\ 2 & 3 & -1 \end{bmatrix} \begin{bmatrix} -2 \\ 3 \\ -1 \end{bmatrix} = [4(2) + 5(3) + 6(-1)] = [17]$$

$$(b) \begin{bmatrix} 2 \\ 3 \\ -1 \end{bmatrix} \begin{bmatrix} 4 & 5 & 6 \end{bmatrix} = \begin{bmatrix} 2(4) & 2(5) & 2(6) \\ 3(4) & 3(5) & 3(6) \\ -1(4) & -1(5) & -1(6) \end{bmatrix} = \begin{bmatrix} 8 & 10 & 12 \\ 12 & 15 & 18 \\ -4 & -5 & -6 \end{bmatrix}$$

$$(c) \begin{bmatrix} 1 & 2 & 3 \end{bmatrix} \begin{bmatrix} 4 & -6 & 9 & 6 \\ 0 & -7 & 10 & 7 \\ 5 & 8 & -11 & -8 \end{bmatrix} \\ = [1(4) + 2(0) + 3(5) \quad 1(-6) + 2(-7) + 3(8) \quad 1(9) + 2(10) + 3(-11) \quad 1(6) + 2(7) + 3(-8)] \\ = [19 \quad 4 \quad -4 \quad -4]$$

$$(d) \begin{bmatrix} 2 & 3 & 4 \\ 1 & 5 & 6 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} = \begin{bmatrix} 2(1) + 3(2) + 4(3) \\ 1(1) + 5(2) + 6(3) \end{bmatrix} = \begin{bmatrix} 20 \\ 29 \end{bmatrix}$$

$$(e) \begin{bmatrix} 1 & 2 & 1 \\ 4 & 0 & 2 \end{bmatrix} \begin{bmatrix} 3 & -4 \\ 1 & 5 \\ -2 & 2 \end{bmatrix} = \begin{bmatrix} 1(3) + 2(1) + 1(-2) & 1(-4) + 2(5) + 1(2) \\ 4(3) + 0(1) + 2(-2) & 4(-4) + 0(5) + 2(2) \end{bmatrix} = \begin{bmatrix} 3 & 8 \\ 8 & -12 \end{bmatrix}$$

4. Let $A = \begin{bmatrix} 2 & -1 & 1 \\ 0 & 1 & 2 \\ 1 & 0 & 1 \end{bmatrix}$. Then

$$A^2 = \begin{bmatrix} 2 & -1 & 1 \\ 0 & 1 & 2 \\ 1 & 0 & 1 \end{bmatrix} \begin{bmatrix} 2 & -1 & 1 \\ 0 & 1 & 2 \\ 1 & 0 & 1 \end{bmatrix} = \begin{bmatrix} 5 & -3 & 1 \\ 2 & 1 & 4 \\ 3 & -1 & 2 \end{bmatrix} \quad \text{and} \quad A^3 = A^2 \cdot A = \begin{bmatrix} 5 & -3 & 1 \\ 2 & 1 & 4 \\ 3 & -1 & 2 \end{bmatrix} \begin{bmatrix} 2 & -1 & 1 \\ 0 & 1 & 2 \\ 1 & 0 & 1 \end{bmatrix} = \begin{bmatrix} 11 & -8 & 0 \\ 8 & -1 & 8 \\ 8 & -4 & 3 \end{bmatrix}$$

The reader will show that $A^3 = A \cdot A^2$ and $A^2 \cdot A^3 = A^3 \cdot A^2$.

5. Show that:

$$(a) \sum_{k=1}^2 a_{ik}(b_{kj} + c_{kj}) = \sum_{k=1}^2 a_{ik}b_{kj} + \sum_{k=1}^2 a_{ik}c_{kj},$$

$$(b) \sum_{i=1}^2 \sum_{j=1}^3 a_{ij} = \sum_{j=1}^3 \sum_{i=1}^2 a_{ij},$$

$$(c) \sum_{k=1}^2 a_{ik} \left(\sum_{h=1}^3 b_{kh}c_{hj} \right) = \sum_{h=1}^3 \left(\sum_{k=1}^2 a_{ik}b_{kh} \right) c_{hj}.$$

$$(a) \sum_{k=1}^2 a_{ik}(b_{kj} + c_{kj}) = a_{i1}(b_{1j} + c_{1j}) + a_{i2}(b_{2j} + c_{2j}) = (a_{i1}b_{1j} + a_{i2}b_{2j}) + (a_{i1}c_{1j} + a_{i2}c_{2j}) \\ = \sum_{k=1}^2 a_{ik}b_{kj} + \sum_{k=1}^2 a_{ik}c_{kj}.$$

$$(b) \sum_{i=1}^2 \sum_{j=1}^3 a_{ij} = \sum_{i=1}^2 (a_{i1} + a_{i2} + a_{i3}) = (a_{11} + a_{12} + a_{13}) + (a_{21} + a_{22} + a_{23}) \\ = (a_{11} + a_{21}) + (a_{12} + a_{22}) + (a_{13} + a_{23}) \\ = \sum_{i=1}^2 a_{i1} + \sum_{i=1}^2 a_{i2} + \sum_{i=1}^2 a_{i3} = \sum_{j=1}^3 \sum_{i=1}^2 a_{ij}.$$

This is simply the statement that in summing all of the elements of a matrix, one may sum first the elements of each row or the elements of each column.

$$(c) \sum_{k=1}^2 a_{ik} \left(\sum_{h=1}^3 b_{kh}c_{hj} \right) = \sum_{k=1}^2 a_{ik}(b_{k1}c_{1j} + b_{k2}c_{2j} + b_{k3}c_{3j}) \\ = a_{i1}(b_{11}c_{1j} + b_{12}c_{2j} + b_{13}c_{3j}) + a_{i2}(b_{21}c_{1j} + b_{22}c_{2j} + b_{23}c_{3j}) \\ = (a_{i1}b_{11} + a_{i2}b_{21})c_{1j} + (a_{i1}b_{12} + a_{i2}b_{22})c_{2j} + (a_{i1}b_{13} + a_{i2}b_{23})c_{3j} \\ = \left(\sum_{k=1}^2 a_{ik}b_{k1} \right) c_{1j} + \left(\sum_{k=1}^2 a_{ik}b_{k2} \right) c_{2j} + \left(\sum_{k=1}^2 a_{ik}b_{k3} \right) c_{3j} \\ = \sum_{h=1}^3 \left(\sum_{k=1}^2 a_{ik}b_{kh} \right) c_{hj}.$$

6. Prove: If $A = [a_{ij}]$ is of order $m \times n$ and if $B = [b_{ij}]$ and $C = [c_{ij}]$ are of order $n \times p$, then $A(B + C) = AB + AC$.

The elements of the i th row of A are $a_{i1}, a_{i2}, \dots, a_{in}$ and the elements of the j th column of $B + C$ are $b_{1j} + c_{1j}, b_{2j} + c_{2j}, \dots, b_{nj} + c_{nj}$. Then the element standing in the i th row and j th column of $A(B + C)$ is $a_{i1}(b_{1j} + c_{1j}) + a_{i2}(b_{2j} + c_{2j}) + \dots + a_{in}(b_{nj} + c_{nj}) = \sum_{k=1}^n a_{ik}(b_{kj} + c_{kj}) = \sum_{k=1}^n a_{ik}b_{kj} + \sum_{k=1}^n a_{ik}c_{kj}$, the sum of the elements standing in the i th row and j th column of AB and AC .

7. Prove: If $A = [a_{ij}]$ is of order $m \times n$, if $B = [b_{ij}]$ is of order $n \times p$, and if $C = [c_{ij}]$ is of order $p \times q$, then $A(BC) = (AB)C$.

The elements of the i th row of A are $a_{i1}, a_{i2}, \dots, a_{in}$ and the elements of the j th column of BC are $\sum_{h=1}^p b_{1h}c_{hj}, \dots, \sum_{h=1}^p b_{nh}c_{hj}$; hence the element standing in the i th row and j th column of $A(BC)$ is $a_{i1} \sum_{h=1}^p b_{1h}c_{hj} + a_{i2} \sum_{h=1}^p b_{2h}c_{hj} + \dots + a_{in} \sum_{h=1}^p b_{nh}c_{hj} = \sum_{k=1}^n a_{ik} \left(\sum_{h=1}^p b_{kh}c_{hj} \right) \\ = \sum_{h=1}^p \left(\sum_{k=1}^n a_{ik}b_{kh} \right) c_{hj} = \left(\sum_{k=1}^n a_{ik}b_{k1} \right) c_{1j} + \left(\sum_{k=1}^n a_{ik}b_{k2} \right) c_{2j} + \dots + \left(\sum_{k=1}^n a_{ik}b_{kp} \right) c_{pj}$

This is the element standing in the i th row and j th column of $(AB)C$; hence, $A(BC) = (AB)C$.

8. Assuming A, B, C, D conformable, show in two ways that $(A + B)(C + D) = AC + AD + BC + BD$.

Using (e) and then (f), $(A + B)(C + D) = (A + B)C + (A + B)D = AC + BC + AD + BD$.

Using (f) and then (e), $(A + B)(C + D) = A(C + D) + B(C + D) = AC + AD + BC + BD = AC + BC + AD + BD$.

$$9. (a) \begin{bmatrix} 1 & 0 & 0 & 1 \\ 0 & 1 & 0 & 2 \\ 0 & 0 & 1 & 3 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \\ 3 & 1 & 2 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} + \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} [3 \ 1 \ 2] = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} + \begin{bmatrix} 3 & 1 & 2 \\ 6 & 2 & 4 \\ 9 & 3 & 6 \end{bmatrix} = \begin{bmatrix} 4 & 1 & 2 \\ 6 & 3 & 4 \\ 9 & 3 & 7 \end{bmatrix}$$

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

$$= \begin{bmatrix} 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 2 & 0 & 0 & 0 & 0 \\ 0 & 0 & 3 & 0 & 0 & 0 \\ 0 & 0 & 0 & 12 & 0 & 0 \\ 0 & 0 & 0 & 0 & 10 & 0 \\ 0 & 0 & 0 & 0 & 0 & 18 \end{bmatrix}$$

$$(c) \begin{bmatrix} 1 & 1 & 0 & 0 & 0 & 0 \\ 2 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 3 & 1 & 2 & 0 \\ 0 & 0 & 1 & 2 & 1 & 0 \\ 0 & 0 & 0 & 1 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 2 & 3 & 4 & 5 & 6 \\ 2 & 3 & 4 & 5 & 6 & 7 \\ 3 & 4 & 5 & 6 & 7 & 8 \\ 4 & 5 & 6 & 7 & 8 & 9 \\ 9 & 8 & 7 & 6 & 5 & 4 \\ 8 & 7 & 6 & 5 & 4 & 1 \end{bmatrix} = \begin{bmatrix} [1 \ 1] [1 \ 2] & [1 \ 1] [3 \ 4 \ 5] & [1 \ 1] [6] \\ [2 \ 1] [2 \ 3] & [2 \ 1] [4 \ 5 \ 6] & [2 \ 1] [7] \\ [3 \ 1 \ 2] [3 \ 4] & [3 \ 1 \ 2] [5 \ 6 \ 7] & [3 \ 1 \ 2] [8] \\ [1 \ 2 \ 1] [4 \ 5] & [1 \ 2 \ 1] [6 \ 7 \ 8] & [1 \ 2 \ 1] [9] \\ [0 \ 1 \ 1] [9 \ 8] & [0 \ 1 \ 1] [7 \ 6 \ 5] & [0 \ 1 \ 1] [4] \\ [1] \cdot [8 \ 7] & [1] \cdot [6 \ 5 \ 4] & [1] \cdot [1] \end{bmatrix}$$

$$= \begin{bmatrix} [3 \ 5] [7 \ 9 \ 11] [13] \\ [4 \ 7] [10 \ 13 \ 16] [19] \\ [31 \ 33] [35 \ 37 \ 39] [41] \\ [20 \ 22] [24 \ 26 \ 28] [30] \\ [13 \ 13] [13 \ 13 \ 13] [13] \\ [8 \ 7] [6 \ 5 \ 4] [1] \end{bmatrix} = \begin{bmatrix} 3 & 5 & 7 & 9 & 11 & 13 \\ 4 & 7 & 10 & 13 & 16 & 19 \\ 31 & 33 & 35 & 37 & 39 & 41 \\ 20 & 22 & 24 & 26 & 28 & 30 \\ 13 & 13 & 13 & 13 & 13 & 13 \\ 8 & 7 & 6 & 5 & 4 & 1 \end{bmatrix}$$

10. Let $\begin{cases} x_1 = a_{11}y_1 + a_{12}y_2 \\ x_2 = a_{21}y_1 + a_{22}y_2 \\ x_3 = a_{31}y_1 + a_{32}y_2 \end{cases}$ be three linear forms in y_1 and y_2 and let $\begin{cases} y_1 = b_{11}z_1 + b_{12}z_2 \\ y_2 = b_{21}z_1 + b_{22}z_2 \end{cases}$ be a

linear transformation of the coordinates (y_1, y_2) into new coordinates (z_1, z_2) . The result of applying the transformation to the given forms is the set of forms

$$\begin{cases} x_1 = (a_{11}b_{11} + a_{12}b_{21})z_1 + (a_{11}b_{12} + a_{12}b_{22})z_2 \\ x_2 = (a_{21}b_{11} + a_{22}b_{21})z_1 + (a_{21}b_{12} + a_{22}b_{22})z_2 \\ x_3 = (a_{31}b_{11} + a_{32}b_{21})z_1 + (a_{31}b_{12} + a_{32}b_{22})z_2 \end{cases}$$

Using matrix notation, we have the three forms $\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \\ a_{31} & a_{32} \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \end{bmatrix}$ and the transformation

$\begin{bmatrix} y_1 \\ y_2 \end{bmatrix} = \begin{bmatrix} b_{11} & b_{12} \\ b_{21} & b_{22} \end{bmatrix} \begin{bmatrix} z_1 \\ z_2 \end{bmatrix}$. The result of applying the transformation is the set of three forms

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \\ a_{31} & a_{32} \end{bmatrix} \begin{bmatrix} b_{11} & b_{12} \\ b_{21} & b_{22} \end{bmatrix} \begin{bmatrix} z_1 \\ z_2 \end{bmatrix}$$

Thus, when a set of m linear forms in n variables with matrix A is subjected to a linear transformation of the variables with matrix B , there results a set of m linear forms with matrix $C = A^p B$

SUPPLEMENTARY PROBLEMS

11. Given $A = \begin{bmatrix} 1 & 2 & -3 \\ 5 & 0 & 2 \\ 1 & -1 & 1 \end{bmatrix}$, $B = \begin{bmatrix} 3 & -1 & 2 \\ 4 & 2 & 5 \\ 2 & 0 & 3 \end{bmatrix}$, and $C = \begin{bmatrix} 4 & 1 & 2 \\ 0 & 3 & 2 \\ 1 & -2 & 3 \end{bmatrix}$,

(a) Compute: $A+B = \begin{bmatrix} 4 & 1 & -1 \\ 9 & 2 & 7 \\ 3 & -1 & 4 \end{bmatrix}$, $A-C = \begin{bmatrix} -3 & 1 & -5 \\ 5 & -3 & 0 \\ 0 & 1 & -2 \end{bmatrix}$

(b) Compute: $-2A = \begin{bmatrix} -2 & -4 & 6 \\ -10 & 0 & -4 \\ -2 & 2 & -2 \end{bmatrix}$, $0 \cdot B = 0$

(c) Verify: $A+(B-C) = (A+B)-C$.

(d) Find the matrix D such that $A+D=B$. Verify that $D=B-A=-(A-B)$.

12. Given $A = \begin{bmatrix} 1 & -1 & 1 \\ -3 & 2 & -1 \\ -2 & 1 & 0 \end{bmatrix}$ and $B = \begin{bmatrix} 1 & 2 & 3 \\ 2 & 4 & 6 \\ 1 & 2 & 3 \end{bmatrix}$, compute $AB=0$ and $BA = \begin{bmatrix} -11 & 6 & -1 \\ -22 & 12 & -2 \\ -11 & 6 & -1 \end{bmatrix}$. Hence, $AB \neq BA$ generally.

13. Given $A = \begin{bmatrix} 1 & -3 & 2 \\ 2 & 1 & -3 \\ 4 & -3 & -1 \end{bmatrix}$, $B = \begin{bmatrix} 1 & 4 & 1 & 0 \\ 2 & 1 & 1 & 1 \\ 1 & -2 & 1 & 2 \end{bmatrix}$, and $C = \begin{bmatrix} 2 & 1 & -1 & -2 \\ 3 & -2 & -1 & -1 \\ 2 & -5 & -1 & 0 \end{bmatrix}$, show that $AB=AC$. Thus, $AB=AC$ does not necessarily imply $B=C$.

14. Given $A = \begin{bmatrix} 1 & 1 & -1 \\ 2 & 0 & 3 \\ 3 & -1 & 2 \end{bmatrix}$, $B = \begin{bmatrix} 1 & 3 \\ 0 & 2 \\ -1 & 4 \end{bmatrix}$, and $C = \begin{bmatrix} 1 & 2 & 3 & -4 \\ 2 & 0 & -2 & 1 \end{bmatrix}$, show that $(AB)C = A(BC)$.

15. Using the matrices of Problem 11, show that $A(B+C) = AB+AC$ and $(A+B)C = AC+BC$.

16. Explain why, in general, $(A \pm B)^2 \neq A^2 \pm 2AB + B^2$ and $A^2 - B^2 \neq (A-B)(A+B)$.

17. Given $A = \begin{bmatrix} 2 & -3 & -5 \\ -1 & 4 & 5 \\ 1 & -3 & -4 \end{bmatrix}$, $B = \begin{bmatrix} -1 & 3 & 5 \\ 1 & -3 & -5 \\ -1 & 3 & 5 \end{bmatrix}$, and $C = \begin{bmatrix} 2 & -2 & -4 \\ -1 & 3 & 4 \\ 1 & -2 & -3 \end{bmatrix}$,

(a) show that $AB=BA=0$, $AC=A$, $CA=C$.

(b) use the results of (a) to show that $ACB=CBA$, $A^2 - B^2 = (A-B)(A+B)$, $(A \pm B)^2 = A^2 + B^2$.

18. Given $A = \begin{bmatrix} i & 0 \\ 0 & i \end{bmatrix}$, where $i^2 = -1$, derive a formula for the positive integral powers of A .

Ans. $A^n = I, A, -I, -A$ according as $n = 4p, 4p+1, 4p+2, 4p+3$, where $I = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$.

19. Show that the product of any two or more matrices of the set $\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix}, \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}, \begin{bmatrix} -1 & 0 \\ 0 & -1 \end{bmatrix}, \begin{bmatrix} i & 0 \\ 0 & -i \end{bmatrix}, \begin{bmatrix} -i & 0 \\ 0 & i \end{bmatrix}, \begin{bmatrix} 0 & -i \\ -i & 0 \end{bmatrix}, \begin{bmatrix} 0 & i \\ i & 0 \end{bmatrix}$ is a matrix of the set.

20. Given the matrices A of order $m \times n$, B of order $n \times p$, and C of order $r \times q$, under what conditions on p, q , and r would the matrices be conformable for finding the products and what is the order of each: (a) ABC , (b) ACB , (c) $A(B+C)$?

Ans. (a) $p=r; m \times q$ (b) $r=n=q; m \times p$ (c) $r=n, p=q; m \times q$

21. Compute AB , given:

$$(a) A = \left[\begin{array}{cc|c} 1 & 0 & 1 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{array} \right] \quad \text{and} \quad B = \left[\begin{array}{cc|c} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 1 & 1 & 0 \end{array} \right] \quad \text{Ans.} \quad \left[\begin{array}{ccc} 2 & 1 & 0 \\ 1 & 2 & 0 \\ 1 & 1 & 0 \end{array} \right]$$

$$(b) A = \left[\begin{array}{cc|c} 1 & 0 & 3 \\ 0 & 1 & 2 \end{array} \right] \quad \text{and} \quad B = \left[\begin{array}{cc} 1 & 0 \\ 0 & 1 \\ -1 & 2 \end{array} \right] \quad \text{Ans.} \quad \left[\begin{array}{cc} -2 & 6 \\ -2 & 5 \end{array} \right]$$

$$(c) A = \left[\begin{array}{cc|cc} 1 & 2 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 2 & 2 \end{array} \right] \quad \text{and} \quad B = \left[\begin{array}{cc|cc} 0 & 0 & 0 & 1 \\ 0 & 0 & 2 & 0 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \end{array} \right] \quad \text{Ans.} \quad \left[\begin{array}{cccc} 0 & 0 & 4 & 1 \\ 0 & 0 & 2 & 0 \\ 0 & 1 & 0 & 0 \\ 2 & 2 & 0 & 0 \end{array} \right]$$

22. Prove: (a) $\text{trace}(A+B) = \text{trace} A + \text{trace} B$, (b) $\text{trace}(kA) = k \text{trace} A$.

23. If $\begin{cases} x_1 = y_1 - 2y_2 + y_3 \\ x_2 = 2y_1 + y_2 - 3y_3 \end{cases}$ and $\begin{cases} y_1 = z_1 + 2z_2 \\ y_2 = 2z_1 - z_2 \\ y_3 = 2z_1 + 3z_2 \end{cases}$, verify $\begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 1 & -2 & 1 \\ 2 & 1 & -3 \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix} = \begin{bmatrix} 1 & -2 & 1 \\ 2 & 1 & -3 \end{bmatrix} \begin{bmatrix} 1 & 2 \\ 2 & -1 \\ 2 & 3 \end{bmatrix} \begin{bmatrix} z_1 \\ z_2 \end{bmatrix}$
 $= \begin{bmatrix} -z_1 + 7z_2 \\ -2z_1 - 6z_2 \end{bmatrix}$

24. If $A = [a_{ij}]$ and $B = [b_{ij}]$ are of order $m \times n$ and if $C = [c_{ij}]$ is of order $n \times p$, show that $(A+B)C = AC + BC$.

25. Let $A = [a_{ij}]$ and $B = [b_{jk}]$, where $(i = 1, 2, \dots, m; j = 1, 2, \dots, p; k = 1, 2, \dots, n)$. Denote by β_j the sum of

the elements of the j th row of B , that is, let $\beta_j = \sum_{k=1}^n b_{jk}$. Show that the element in the i th row of $A \cdot \begin{bmatrix} \beta_1 \\ \beta_2 \\ \vdots \\ \beta_p \end{bmatrix}$

is the sum of the elements lying in the i th row of AB . Use this procedure to check the products formed in Problems 12 and 13.

26. A relation (such as parallelism, congruency) between mathematical entities possessing the following properties:

- (i) Determination Either a is in the relation to b or a is not in the relation to b .
- (ii) Reflexivity a is in the relation to a , for all a .
- (iii) Symmetry If a is in the relation to b then b is in the relation to a .
- (iv) Transitivity If a is in the relation to b and b is in the relation to c then a is in the relation to c .

is called an **equivalence relation**.

Show that the parallelism of lines, similarity of triangles, and equality of matrices are equivalence relations. Show that perpendicularity of lines is not an equivalence relation.

27. Show that conformability for addition of matrices is an equivalence relation while conformability for multiplication is not.

28. Prove: If A, B, C are matrices such that $AC = CA$ and $BC = CB$, then $(AB \pm BA)C = C(AB \pm BA)$.

Chapter 2

Some Types of Matrices

THE IDENTITY MATRIX. A square matrix A whose elements $a_{ij} = 0$ for $i > j$ is called **upper triangular**; a square matrix A whose elements $a_{ij} = 0$ for $i < j$ is called **lower triangular**. Thus

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} & \dots & a_{1n} \\ 0 & a_{22} & a_{23} & \dots & a_{2n} \\ 0 & 0 & a_{33} & \dots & a_{3n} \\ \dots & \dots & \dots & \dots & \dots \\ 0 & 0 & 0 & \dots & a_{nn} \end{bmatrix}$$

is upper triangular and

$$\begin{bmatrix} a_{11} & 0 & 0 & \dots & 0 \\ a_{21} & a_{22} & 0 & \dots & 0 \\ a_{31} & a_{32} & a_{33} & \dots & 0 \\ \dots & \dots & \dots & \dots & \dots \\ a_{n1} & a_{n2} & a_{n3} & \dots & a_{nn} \end{bmatrix}$$

is lower triangular.

The matrix $D = \begin{bmatrix} a_{11} & 0 & 0 & \dots & 0 \\ 0 & a_{22} & 0 & \dots & 0 \\ 0 & 0 & a_{33} & \dots & 0 \\ \dots & \dots & \dots & \dots & \dots \\ 0 & 0 & 0 & \dots & a_{nn} \end{bmatrix}$, which is both upper and lower triangular, is called a **diagonal matrix**.

It will frequently be written as

$$D = \text{diag}(a_{11}, a_{22}, a_{33}, \dots, a_{nn})$$

See Problem 1.

If in the diagonal matrix D above, $a_{11} = a_{22} = \dots = a_{nn} = k$, D is called a **scalar matrix**; if, in addition, $k = 1$, the matrix is called the **identity matrix** and is denoted by I_n . For example

$$I_2 = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

and

$$I_3 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

When the order is evident or immaterial, an identity matrix will be denoted by I . Clearly, $I_n + I_n + \dots$ to p terms $= p \cdot I_n = \text{diag}(p, p, p, \dots, p)$ and $I^p = I \cdot I \dots$ to p factors $= I$. Identity matrices have some of the properties of the integer 1. For example, if $A = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{bmatrix}$ then $I_2 \cdot A =$

$= A \cdot I_3 = I_2 A I_3 = A$, as the reader may readily show.

SPECIAL SQUARE MATRICES. If A and B are square matrices such that $AB = BA$, then A and B are called **commutative** or are said to **commute**. It is a simple matter to show that if A is any n -square matrix, it commutes with itself and also with I_n .

See Problem 2.

If A and B are such that $AB = -BA$, the matrices A and B are said to **anti-commute**.

A matrix A for which $A^{k+1} = A$, where k is a positive integer, is called **periodic**. If k is the least positive integer for which $A^{k+1} = A$, then A is said to be of **period k** .

If $k = 1$, so that $A^2 = A$, then A is called **idempotent**.

See Problems 3-4.

A matrix A for which $A^p = 0$, where p is a positive integer, is called **nilpotent**. If p is the least positive integer for which $A^p = 0$, then A is said to be nilpotent of **index p** .

See Problems 5-6.

THE INVERSE OF A MATRIX. If A and B are square matrices such that $AB = BA = I$, then B is called the **inverse of A** and we write $B = A^{-1}$ (B equals A inverse). The matrix B also has A as its inverse and we may write $A = B^{-1}$.

Example 1. Since $\begin{bmatrix} 1 & 2 & 3 \\ 1 & 3 & 3 \\ 1 & 2 & 4 \end{bmatrix} \begin{bmatrix} 6 & -2 & -3 \\ -1 & 1 & 0 \\ -1 & 0 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} = I$, each matrix in the product is the inverse of the other.

We shall find later (Chapter 7) that not every square matrix has an inverse. We can show here, however, that if A has an inverse then that inverse is unique.

See Problem 7.

If A and B are square matrices of the same order with inverses A^{-1} and B^{-1} respectively, then $(AB)^{-1} = B^{-1} \cdot A^{-1}$, that is,

I. The inverse of the product of two matrices, having inverses, is the product in reverse order of these inverses.

See Problem 8.

A matrix A such that $A^2 = I$ is called **involutory**. An identity matrix, for example, is involutory. An involutory matrix is its own inverse.

See Problem 9.

THE TRANSPOSE OF A MATRIX. The matrix of order $n \times m$ obtained by interchanging the rows and columns of an $m \times n$ matrix A is called the **transpose of A** and is denoted by A' (A transpose). For

example, the transpose of $A = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{bmatrix}$ is $A' = \begin{bmatrix} 1 & 4 \\ 2 & 5 \\ 3 & 6 \end{bmatrix}$. Note that the element a_{ij} in the i th row and j th column of A stands in the j th row and i th column of A' .

If A' and B' are transposes respectively of A and B , and if k is a scalar, we have immediately

$$(a) (A')' = A \quad \text{and} \quad (b) (kA)' = kA'$$

In Problems 10 and 11, we prove:

II. The transpose of the sum of two matrices is the sum of their transposes, i.e.,

$$(A+B)' = A' + B'$$

and

III. The transpose of the product of two matrices is the product in reverse order of their transposes, i.e.,

$$(AB)' = B' \cdot A'$$

See Problems 10-12.

SYMMETRIC MATRICES. A square matrix A such that $A' = A$ is called **symmetric**. Thus, a square matrix $A = [a_{ij}]$ is symmetric provided $a_{ij} = a_{ji}$, for all values of i and j . For example,

$$A = \begin{bmatrix} 1 & 2 & 3 \\ 2 & 4 & -5 \\ 3 & -5 & 6 \end{bmatrix} \text{ is symmetric and so also is } kA \text{ for any scalar } k.$$

In Problem 13, we prove

IV. If A is an n -square matrix, then $A + A'$ is symmetric.

A square matrix A such that $A' = -A$ is called **skew-symmetric**. Thus, a square matrix A is skew-symmetric provided $a_{ij} = -a_{ji}$ for all values of i and j . Clearly, the diagonal elements are

$$\text{zeros. For example, } A = \begin{bmatrix} 0 & -2 & 3 \\ 2 & 0 & 4 \\ -3 & -4 & 0 \end{bmatrix} \text{ is skew-symmetric and so also is } kA \text{ for any scalar } k.$$

With only minor changes in Problem 13, we can prove

V. If A is any n -square matrix, then $A - A'$ is skew-symmetric.

From Theorems IV and V follows

VI. Every square matrix A can be written as the sum of a symmetric matrix $B = \frac{1}{2}(A + A')$ and a skew-symmetric matrix $C = \frac{1}{2}(A - A')$. See Problems 14-15.

THE CONJUGATE OF A MATRIX. Let a and b be real numbers and let $i = \sqrt{-1}$; then, $z = a + bi$ is called a **complex number**. The complex numbers $a + bi$ and $a - bi$ are called **conjugates**, each being the **conjugate** of the other. If $z = a + bi$, its conjugate is denoted by $\bar{z} = a - bi$.

If $z_1 = a + bi$ and $z_2 = \bar{z}_1 = a - bi$, then $\bar{z}_2 = \overline{\bar{z}_1} = a + bi = z_1$, that is, the conjugate of the conjugate of a complex number z is z itself.

If $z_1 = a + bi$ and $z_2 = c + di$, then

$$(i) \quad z_1 + z_2 = (a+c) + (b+d)i \quad \text{and} \quad \overline{z_1 + z_2} = (a+c) - (b+d)i = (a-bi) + (c-di) = \bar{z}_1 + \bar{z}_2,$$

that is, the conjugate of the sum of two complex numbers is the sum of their conjugates.

$$(ii) \quad z_1 \cdot z_2 = (ac - bd) + (ad + bc)i \quad \text{and} \quad \overline{z_1 \cdot z_2} = (ac - bd) - (ad + bc)i = (a - bi)(c - di) = \bar{z}_1 \cdot \bar{z}_2,$$

that is, the conjugate of the product of two complex numbers is the product of their conjugates.

When A is a matrix having complex numbers as elements, the matrix obtained from A by replacing each element by its conjugate is called the **conjugate** of A and is denoted by \bar{A} (A conjugate).

Example 2. When $A = \begin{bmatrix} 1+2i & i \\ 3 & 2-3i \end{bmatrix}$ then $\bar{A} = \begin{bmatrix} 1-2i & -i \\ 3 & 2+3i \end{bmatrix}$.

If \bar{A} and \bar{B} are the conjugates of the matrices A and B and if k is any scalar, we have readily

$$(c) \quad \overline{\bar{A}} = A \quad \text{and} \quad (d) \quad \overline{kA} = \bar{k} \cdot \bar{A}$$

Using (i) and (ii) above, we may prove

VII. The conjugate of the sum of two matrices is the sum of their conjugates, i.e., $\overline{A+B} = \overline{A} + \overline{B}$.

VIII. The conjugate of the product of two matrices is the product, in the same order, of their conjugates, i.e., $\overline{AB} = \overline{A}\overline{B}$.

The transpose of \overline{A} is denoted by \overline{A}' (A conjugate transpose). It is sometimes written as A^* . We have

IX. The transpose of the conjugate of A is equal to the conjugate of the transpose of A , i.e., $(\overline{A})' = \overline{A'}$.

Example 3. From Example 2

$$(\overline{A})' = \begin{bmatrix} 1-2i & 3 \\ -i & 2+3i \end{bmatrix} \quad \text{while} \quad A' = \begin{bmatrix} 1+2i & 3 \\ i & 2-3i \end{bmatrix} \quad \text{and} \quad \overline{A'} = \begin{bmatrix} 1-2i & 3 \\ -i & 2+3i \end{bmatrix} = (\overline{A})'$$

HERMITIAN MATRICES. A square matrix $A = [a_{ij}]$ such that $\overline{A'} = A$ is called **Hermitian**. Thus, A is Hermitian provided $a_{ij} = \overline{a_{ji}}$ for all values of i and j . Clearly, the diagonal elements of an Hermitian matrix are real numbers.

Example 4. The matrix $A = \begin{bmatrix} 1 & 1-i & 2 \\ 1+i & 3 & i \\ 2 & -i & 0 \end{bmatrix}$ is Hermitian.

Is kA Hermitian if k is any real number? any complex number?

A square matrix $A = [a_{ij}]$ such that $\overline{A'} = -A$ is called **skew-Hermitian**. Thus, A is skew-Hermitian provided $a_{ij} = -\overline{a_{ji}}$ for all values of i and j . Clearly, the diagonal elements of a skew-Hermitian matrix are either zeros or pure imaginaries.

Example 5. The matrix $A = \begin{bmatrix} i & 1-i & 2 \\ -1-i & 3i & i \\ -2 & i & 0 \end{bmatrix}$ is skew-Hermitian. Is kA skew-Hermitian if k is any real number? any complex number? any pure imaginary?

By making minor changes in Problem 13, we may prove

X. If A is an n -square matrix then $A + \overline{A'}$ is Hermitian and $A - \overline{A'}$ is skew-Hermitian.

From Theorem X follows

XI. Every square matrix A with complex elements can be written as the sum of an Hermitian matrix $B = \frac{1}{2}(A + \overline{A'})$ and a skew-Hermitian matrix $C = \frac{1}{2}(A - \overline{A'})$.

DIRECT SUM. Let A_1, A_2, \dots, A_s be square matrices of respective orders m_1, m_2, \dots, m_s . The generalization

$$A = \begin{bmatrix} A_1 & 0 & \dots & 0 \\ 0 & A_2 & \dots & 0 \\ \dots & \dots & \dots & \dots \\ 0 & 0 & \dots & A_s \end{bmatrix} = \text{diag}(A_1, A_2, \dots, A_s)$$

of the diagonal matrix is called the **direct sum** of the A_i .

Example 6. Let $A_1 = [2]$, $A_2 = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$, and $A_3 = \begin{bmatrix} 1 & 2 & -1 \\ 2 & 0 & 3 \\ 4 & 1 & -2 \end{bmatrix}$.

The direct sum of A_1, A_2, A_3 is $\text{diag}(A_1, A_2, A_3) = \begin{bmatrix} 2 & 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 2 & 0 & 0 & 0 \\ 0 & 3 & 4 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 2 & -1 \\ 0 & 0 & 0 & 2 & 0 & 3 \\ 0 & 0 & 0 & 4 & 1 & -2 \end{bmatrix}$

Problem 9(b), Chapter 1, illustrates

XII. If $A = \text{diag}(A_1, A_2, \dots, A_s)$ and $B = \text{diag}(B_1, B_2, \dots, B_s)$, where A_i and B_i have the same order for $(i = 1, 2, \dots, s)$, then $AB = \text{diag}(A_1B_1, A_2B_2, \dots, A_sB_s)$.

SOLVED PROBLEMS

1. Since $\begin{bmatrix} a_{11} & 0 & \dots & 0 \\ 0 & a_{22} & \dots & 0 \\ \dots & \dots & \dots & \dots \\ 0 & 0 & \dots & a_{mm} \end{bmatrix} \begin{bmatrix} b_{11} & b_{12} & \dots & b_{1n} \\ b_{21} & b_{22} & \dots & b_{2n} \\ \dots & \dots & \dots & \dots \\ b_{m1} & b_{m2} & \dots & b_{mn} \end{bmatrix} = \begin{bmatrix} a_{11}b_{11} & a_{11}b_{12} & \dots & a_{11}b_{1n} \\ a_{22}b_{21} & a_{22}b_{22} & \dots & a_{22}b_{2n} \\ \dots & \dots & \dots & \dots \\ a_{mm}b_{m1} & a_{mm}b_{m2} & \dots & a_{mm}b_{mn} \end{bmatrix}$, the product AB of

an m -square diagonal matrix $A = \text{diag}(a_{11}, a_{22}, \dots, a_{mm})$ and any $m \times n$ matrix B is obtained by multiplying the first row of B by a_{11} , the second row of B by a_{22} , and so on.

2. Show that the matrices $\begin{bmatrix} a & b \\ b & a \end{bmatrix}$ and $\begin{bmatrix} c & d \\ d & c \end{bmatrix}$ commute for all values of a, b, c, d .

This follows from $\begin{bmatrix} a & b \\ b & a \end{bmatrix} \begin{bmatrix} c & d \\ d & c \end{bmatrix} = \begin{bmatrix} ac+bd & ad+bc \\ bc+ad & bd+ac \end{bmatrix} = \begin{bmatrix} c & d \\ d & c \end{bmatrix} \begin{bmatrix} a & b \\ b & a \end{bmatrix}$.

3. Show that $\begin{bmatrix} 2 & -2 & -4 \\ -1 & 3 & 4 \\ 1 & -2 & -3 \end{bmatrix}$ is idempotent.

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

4. Show that if $AB = A$ and $BA = B$, then A and B are idempotent.

$ABA = (AB)A = A \cdot A = A^2$ and $ABA = A(BA) = AB = A$; then $A^2 = A$ and A is idempotent. Use BAB to show that B is idempotent.

5. Show that $A = \begin{bmatrix} 1 & 1 & 3 \\ 5 & 2 & 6 \\ -2 & -1 & -3 \end{bmatrix}$ is nilpotent of order 3.

$$A^2 = \begin{bmatrix} 1 & 1 & 3 \\ 5 & 2 & 6 \\ -2 & -1 & -3 \end{bmatrix} \begin{bmatrix} 1 & 1 & 3 \\ 5 & 2 & 6 \\ -2 & -1 & -3 \end{bmatrix} = \begin{bmatrix} 0 & 0 & 0 \\ 3 & 3 & 9 \\ -1 & -1 & -3 \end{bmatrix} \quad \text{and} \quad A^3 = A^2 \cdot A = \begin{bmatrix} 0 & 0 & 0 \\ 3 & 3 & 9 \\ -1 & -1 & -3 \end{bmatrix} \begin{bmatrix} 1 & 1 & 3 \\ 5 & 2 & 6 \\ -2 & -1 & -3 \end{bmatrix} = 0$$

6. If A is nilpotent of index 2, show that $A(I \pm A)^n = A$ for n any positive integer.

Since $A^2 = 0$, $A^3 = A^4 = \dots = A^n = 0$. Then $A(I \pm A)^n = A(I \pm nA) = A \pm nA^2 = A$.

7. Let A, B, C be square matrices such that $AB = I$ and $CA = I$. Then $(CA)B = C(AB)$ so that $B = C$. Thus, $B = C = A^{-1}$ is the unique inverse of A . (What is B^{-1} ?)

8. Prove: $(AB)^{-1} = B^{-1} \cdot A^{-1}$.

By definition $(AB)^{-1}(AB) = (AB)(AB)^{-1} = I$. Now

$$(B^{-1} \cdot A^{-1})AB = B^{-1}(A^{-1} \cdot A)B = B^{-1} \cdot I \cdot B = B^{-1} \cdot B = I$$

and

$$AB(B^{-1} \cdot A^{-1}) = A(B \cdot B^{-1})A^{-1} = A \cdot I \cdot A^{-1} = I$$

By Problem 7, $(AB)^{-1}$ is unique; hence, $(AB)^{-1} = B^{-1} \cdot A^{-1}$.

9. Prove: A matrix A is involutory if and only if $(I - A)(I + A) = 0$.

Suppose $(I - A)(I + A) = I - A^2 = 0$; then $A^2 = I$ and A is involutory.

Suppose A is involutory; then $A^2 = I$ and $(I - A)(I + A) = I - A^2 = I - I = 0$.

10. Prove: $(A + B)' = A' + B'$.

Let $A = [a_{ij}]$ and $B = [b_{ij}]$. We need only check that the element in the i th row and j th column of A' , B' , and $(A + B)'$ are respectively a_{ji} , b_{ji} , and $a_{ji} + b_{ji}$.

11. Prove: $(AB)' = B'A'$.

Let $A = [a_{ij}]$ be of order $m \times n$, $B = [b_{ij}]$ be of order $n \times p$; then $C = AB = [c_{ij}]$ is of order $m \times p$. The element standing in the i th row and j th column of AB is $c_{ij} = \sum_{k=1}^n a_{ik} \cdot b_{kj}$ and this is also the element standing in the j th row and i th column of $(AB)'$.

The elements of the j th row of B' are $b_{1j}, b_{2j}, \dots, b_{nj}$ and the elements of the i th column of A' are $a_{i1}, a_{i2}, \dots, a_{in}$. Then the element in the j th row and i th column of $B'A'$ is

$$\sum_{k=1}^n b_{kj} \cdot a_{ik} = \sum_{k=1}^n a_{ik} \cdot b_{kj} = c_{ij}$$

Thus, $(AB)' = B'A'$.

12. Prove: $(ABC)' = C'B'A'$.

Write $ABC = (AB)C$. Then, by Problem 11, $(ABC)' = \{(AB)C\}' = C'(AB)' = C'B'A'$.

13. Show that if $A = [a_{ij}]$ is n -square, then $B = [b_{ij}] = A + A'$ is symmetric.

First Proof.

The element in the i th row and j th column of A is a_{ij} and the corresponding element of A' is a_{ji} ; hence, $b_{ij} = a_{ij} + a_{ji}$. The element in the j th row and i th column of A is a_{ji} and the corresponding element of A' is a_{ij} ; hence, $b_{ji} = a_{ji} + a_{ij}$. Thus, $b_{ij} = b_{ji}$ and B is symmetric.

Second Proof.

By Problem 10, $(A + A')' = A' + (A')' = A' + A = A + A'$ and $(A + A')$ is symmetric.

14. Prove: If A and B are n -square symmetric matrices then AB is symmetric if and only if A and B commute.

Suppose A and B commute so that $AB = BA$. Then $(AB)' = B'A' = BA = AB$ and AB is symmetric.

Suppose AB is symmetric so that $(AB)' = AB$. Now $(AB)' = B'A' = BA$; hence, $AB = BA$ and the matrices A and B commute.

15. Prove: If the m -square matrix A is symmetric (skew-symmetric) and if P is of order $m \times n$ then $B = P'AP$ is symmetric (skew-symmetric).

If A is symmetric then (see Problem 12) $B' = (P'AP)' = P'A'(P')' = P'A'P = P'AP$ and B is symmetric.

If A is skew-symmetric then $B' = (P'AP)' = -P'AP$ and B is skew-symmetric.

16. Prove: If A and B are n -square matrices then A and B commute if and only if $A - kI$ and $B - kI$ commute for every scalar k .

Suppose A and B commute; then $AB = BA$ and

$$\begin{aligned} (A - kI)(B - kI) &= AB - k(A + B) + k^2I \\ &= BA - k(A + B) + k^2I = (B - kI)(A - kI) \end{aligned}$$

Thus, $A - kI$ and $B - kI$ commute.

Suppose $A - kI$ and $B - kI$ commute; then

$$\begin{aligned} (A - kI)(B - kI) &= AB - k(A + B) + k^2I \\ &= BA - k(A + B) + k^2I = (B - kI)(A - kI) \end{aligned}$$

$AB = BA$, and A and B commute.

SUPPLEMENTARY PROBLEMS

17. Show that the product of two upper (lower) triangular matrices is upper (lower) triangular.
18. Derive a rule for forming the product BA of an $m \times n$ matrix B and $A = \text{diag}(a_{11}, a_{22}, \dots, a_{nn})$.
Hint. See Problem 1.
19. Show that the scalar matrix with diagonal element k can be written as kI and that $kA = kIA = \text{diag}(k, k, \dots, k) A$, where the order of I is the row order of A .
20. If A is n -square, show that $A^p \cdot A^q = A^q \cdot A^p$ where p and q are positive integers.
21. (a) Show that $A = \begin{bmatrix} 2 & -3 & -5 \\ -1 & 4 & 5 \\ 1 & -3 & -4 \end{bmatrix}$ and $B = \begin{bmatrix} -1 & 3 & 5 \\ 1 & -3 & -5 \\ -1 & 3 & 5 \end{bmatrix}$ are idempotent.
(b) Using A and B , show that the converse of Problem 4 does not hold.
22. If A is idempotent, show that $B = I - A$ is idempotent and that $AB = BA = 0$.
23. (a) If $A = \begin{bmatrix} 1 & 2 & 2 \\ 2 & 1 & 2 \\ 2 & 2 & 1 \end{bmatrix}$, show that $A^2 - 4A - 5I = 0$.
(b) If $A = \begin{bmatrix} 2 & 1 & 3 \\ 1 & -1 & 2 \\ 1 & 2 & 1 \end{bmatrix}$, show that $A^3 - 2A^2 - 9A = 0$, but $A^2 - 2A - 9I \neq 0$.
24. Show that $\begin{bmatrix} -1 & -1 & -1 \\ 0 & -1 & 0 \\ 0 & 0 & 1 \end{bmatrix}^2 = \begin{bmatrix} 0 & 1 & 0 \\ -1 & -1 & -1 \\ 0 & 0 & 1 \end{bmatrix}^3 = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ -1 & -1 & -1 \end{bmatrix}^4 = I$.
25. Show that $A = \begin{bmatrix} 1 & -2 & -6 \\ -3 & 2 & -9 \\ 2 & 0 & -3 \end{bmatrix}$ is periodic, of period 2.
26. Show that $\begin{bmatrix} 1 & -3 & -4 \\ -1 & 3 & 4 \\ 1 & -3 & -4 \end{bmatrix}$ is nilpotent.
27. Show that (a) $A = \begin{bmatrix} 1 & 2 & 3 \\ 3 & 2 & 0 \\ -1 & -1 & -1 \end{bmatrix}$ and $B = \begin{bmatrix} -2 & -1 & -6 \\ 3 & 2 & 9 \\ -1 & -1 & -4 \end{bmatrix}$ commute,
(b) $A = \begin{bmatrix} 1 & 1 & 2 \\ 2 & 3 & 1 \\ -1 & 2 & 4 \end{bmatrix}$ and $B = \begin{bmatrix} 2/3 & 0 & -1/3 \\ -3/5 & 2/5 & 1/5 \\ 7/15 & -1/5 & 1/15 \end{bmatrix}$ commute.
28. Show that $A = \begin{bmatrix} 1 & -1 \\ 2 & -1 \end{bmatrix}$ and $B = \begin{bmatrix} 1 & 1 \\ 4 & -1 \end{bmatrix}$ anti-commute and $(A+B)^2 = A^2 + B^2$.
29. Show that each of $\begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$, $\begin{bmatrix} 0 & -i \\ i & 0 \end{bmatrix}$, $\begin{bmatrix} i & 0 \\ 0 & -i \end{bmatrix}$ anti-commutes with the others.
30. Prove: The only matrices which commute with every n -square matrix are the n -square scalar matrices.
31. (a) Find all matrices which commute with $\text{diag}(1, 2, 3)$.
(b) Find all matrices which commute with $\text{diag}(a_{11}, a_{22}, \dots, a_{nn})$.
Ans. (a) $\text{diag}(a, b, c)$ where a, b, c are arbitrary.

32. Show that (a) $\begin{bmatrix} 1 & 2 & 3 \\ 2 & 5 & 7 \\ -2 & -4 & -5 \end{bmatrix}$ is the inverse of $\begin{bmatrix} 3 & -2 & -1 \\ -4 & 1 & -1 \\ 2 & 0 & 1 \end{bmatrix}$

(b) $\begin{bmatrix} 1 & 0 & 0 & 0 \\ 2 & 1 & 0 & 0 \\ 4 & 2 & 1 & 0 \\ -2 & 3 & 1 & 1 \end{bmatrix}$ is the inverse of $\begin{bmatrix} 1 & 0 & 0 & 0 \\ -2 & 1 & 0 & 0 \\ 0 & -2 & 1 & 0 \\ 8 & -1 & -1 & 1 \end{bmatrix}$.

33. Set $\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \begin{bmatrix} a & b \\ c & d \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$ to find the inverse of $\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$. Ans. $\begin{bmatrix} -2 & 1 \\ 3/2 & -1/2 \end{bmatrix}$

34. Show that the inverse of a diagonal matrix A , all of whose diagonal elements are different from zero, is a diagonal matrix whose diagonal elements are the inverses of those of A and in the same order. Thus, the inverse of I_n is I_n .

35. Show that $A = \begin{bmatrix} 0 & 1 & -1 \\ 4 & -3 & 4 \\ 3 & -3 & 4 \end{bmatrix}$ and $B = \begin{bmatrix} 4 & 3 & 3 \\ -1 & 0 & -1 \\ -4 & -4 & -3 \end{bmatrix}$ are involutory.

36. Let $A = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ a & b & -1 & 0 \\ c & d & 0 & -1 \end{bmatrix} = \begin{bmatrix} I_2 & 0 \\ A_{21} & -I_2 \end{bmatrix}$ by partitioning. Show that $A^2 = \begin{bmatrix} I_2 & 0 \\ 0 & I_2 \end{bmatrix} = I_4$.

37. Prove: (a) $(A')' = A$, (b) $(kA)' = kA'$, (c) $(A^p)' = (A')^p$ for p a positive integer.

38. Prove: $(ABC)^{-1} = C^{-1}B^{-1}A^{-1}$. *Hint.* Write $ABC = (AB)C$.

39. Prove: (a) $(A^{-1})^{-1} = A$, (b) $(kA)^{-1} = \frac{1}{k}A^{-1}$, (c) $(A^p)^{-1} = (A^{-1})^p$ for p a positive integer.

40. Show that every real symmetric matrix is Hermitian.

41. Prove: (a) $\overline{\overline{A}} = A$, (b) $\overline{A+B} = \overline{A} + \overline{B}$, (c) $\overline{kA} = k\overline{A}$, (d) $\overline{AB} = \overline{A}\overline{B}$.

42. Show: (a) $A = \begin{bmatrix} 1 & 1+i & 2+3i \\ 1-i & 2 & -i \\ 2-3i & i & 0 \end{bmatrix}$ is Hermitian,

(b) $B = \begin{bmatrix} i & 1+i & 2-3i \\ -1+i & 2i & 1 \\ -2-3i & -1 & 0 \end{bmatrix}$ is skew-Hermitian,

(c) iB is Hermitian,

(d) \overline{A} is Hermitian and \overline{B} is skew-Hermitian.

43. If A is n -square, show that (a) AA' and $A'A$ are symmetric, (b) $A+\overline{A'}$, $A\overline{A'}$, and $\overline{A'}A$ are Hermitian.

44. Prove: If H is Hermitian and A is any conformable matrix then $(\overline{A'})'HA$ is Hermitian.

45. Prove: Every Hermitian matrix A can be written as $B+iC$ where B is real and symmetric and C is real and skew-symmetric.

46. Prove: (a) Every skew-Hermitian matrix A can be written as $A = B+iC$ where B is real and skew-symmetric and C is real and symmetric. (b) $A'A$ is real if and only if B and C anti-commute.

47. Prove: If A and B commute so also do A^{-1} and B^{-1} , A' and B' , and $\overline{A'}$ and $\overline{B'}$.

48. Show that for m and n positive integers, A^m and B^n commute if A and B commute.

49. Show (a)
$$\begin{bmatrix} \lambda & 1 \\ 0 & \lambda \end{bmatrix}^n = \begin{bmatrix} \lambda^n & n\lambda^{n-1} \\ 0 & \lambda^n \end{bmatrix}$$
 (b)
$$\begin{bmatrix} \lambda & 1 & 0 \\ 0 & \lambda & 1 \\ 0 & 0 & \lambda \end{bmatrix}^n = \begin{bmatrix} \lambda^n & n\lambda^{n-1} & \frac{1}{2}n(n-1)\lambda^{n-2} \\ 0 & \lambda^n & n\lambda^{n-1} \\ 0 & 0 & \lambda^n \end{bmatrix}$$

50. Prove: If A is symmetric or skew-symmetric then $AA' = A'A$ and A^2 are symmetric.
51. Prove: If A is symmetric so also is $aA^p + bA^{p-1} + \dots + gI$ where a, b, \dots, g are scalars and p is a positive integer.
52. Prove: Every square matrix A can be written as $A = B + C$ where B is Hermitian and C is skew-Hermitian.
53. Prove: If A is real and skew-symmetric or if A is complex and skew-Hermitian then $\pm iA$ are Hermitian.
54. Show that the theorem of Problem 52 can be stated:
Every square matrix A can be written as $A = B + iC$ where B and C are Hermitian.
55. Prove: If A and B are such that $AB = A$ and $BA = B$ then (a) $B'A' = A'$ and $A'B' = B'$, (b) A' and B' are idempotent, (c) $A = B = I$ if A has an inverse.
56. If A is involutory, show that $\frac{1}{2}(I+A)$ and $\frac{1}{2}(I-A)$ are idempotent and $\frac{1}{2}(I+A) \cdot \frac{1}{2}(I-A) = 0$.
57. If the n -square matrix A has an inverse A^{-1} , show:
(a) $(A^{-1})' = (A')^{-1}$, (b) $(\overline{A})^{-1} = \overline{A^{-1}}$, (c) $(\overline{A'})^{-1} = \overline{(A^{-1})'}$
Hint. (a) From the transpose of $AA^{-1} = I$, obtain $(A^{-1})'$ as the inverse of A' .
58. Find all matrices which commute with (a) $\text{diag}(1, 1, 2, 3)$, (b) $\text{diag}(1, 1, 2, 2)$.
Ans: (a) $\text{diag}(A, b, c)$, (b) $\text{diag}(A, B)$ where A and B are 2-square matrices with arbitrary elements and b, c are scalars.
59. If A_1, A_2, \dots, A_s are scalar matrices of respective orders m_1, m_2, \dots, m_s , find all matrices which commute with $\text{diag}(A_1, A_2, \dots, A_s)$.
Ans. $\text{diag}(B_1, B_2, \dots, B_s)$ where B_1, B_2, \dots, B_s are of respective orders m_1, m_2, \dots, m_s with arbitrary elements.
60. If $AB = 0$, where A and B are non-zero n -square matrices, then A and B are called **divisors of zero**. Show that the matrices A and B of Problem 21 are divisors of zero.
61. If $A = \text{diag}(A_1, A_2, \dots, A_s)$ and $B = \text{diag}(B_1, B_2, \dots, B_s)$ where A_i and B_i are of the same order, ($i = 1, 2, \dots, s$), show that
(a) $A + B = \text{diag}(A_1 + B_1, A_2 + B_2, \dots, A_s + B_s)$
(b) $AB = \text{diag}(A_1 B_1, A_2 B_2, \dots, A_s B_s)$
(c) $\text{trace } AB = \text{trace } A_1 B_1 + \text{trace } A_2 B_2 + \dots + \text{trace } A_s B_s$.
62. Prove: If A and B are n -square skew-symmetric matrices then AB is symmetric if and only if A and B commute.
63. Prove: If A is n -square and $B = rA + sI$, where r and s are scalars, then A and B commute.
64. Let A and B be n -square matrices and let r_1, r_2, s_1, s_2 be scalars such that $r_1 s_2 \neq r_2 s_1$. Prove that $C_1 = r_1 A + s_1 B$, $C_2 = r_2 A + s_2 B$ commute if and only if A and B commute.
65. Show that the n -square matrix A will not have an inverse when (a) A has a row (column) of zero elements or (b) A has two identical rows (columns) or (c) A has a row (column) which is the sum of two other rows (columns).
66. If A and B are n -square matrices and A has an inverse, show that
$$(A+B)A^{-1}(A-B) = (A-B)A^{-1}(A+B)$$

Chapter 3

Determinant of a Square Matrix

PERMUTATIONS. Consider the $3! = 6$ permutations of the integers 1, 2, 3 taken together

$$(3.1) \quad 123 \quad 132 \quad 213 \quad 231 \quad 312 \quad 321$$

and eight of the $4! = 24$ permutations of the integers 1, 2, 3, 4 taken together

$$(3.2) \quad \begin{array}{cccc} 1234 & 2134 & 3124 & 4123 \\ 1324 & 2314 & 3214 & 4213 \end{array}$$

If in a given permutation a larger integer precedes a smaller one, we say that there is an **inversion**. If in a given permutation the number of inversions is even (odd), the permutation is called **even (odd)**. For example, in (3.1) the permutation 123 is even since there is no inversion, the permutation 132 is odd since in it 3 precedes 2, the permutation 312 is even since in it 3 precedes 1 and 3 precedes 2. In (3.2) the permutation 4213 is even since in it 4 precedes 2, 4 precedes 1, 4 precedes 3, and 2 precedes 1.

THE DETERMINANT OF A SQUARE MATRIX. Consider the n -square matrix

$$(3.3) \quad A = \begin{bmatrix} a_{11} & a_{12} & a_{13} & \cdots & a_{1n} \\ a_{21} & a_{22} & a_{23} & \cdots & a_{2n} \\ \cdots & \cdots & \cdots & \cdots & \cdots \\ a_{n1} & a_{n2} & a_{n3} & \cdots & a_{nn} \end{bmatrix}$$

and a product

$$(3.4) \quad a_{1j_1} a_{2j_2} a_{3j_3} \cdots a_{nj_n}$$

of n of its elements, selected so that one and only one element comes from any row and one and only one element comes from any column. In (3.4), as a matter of convenience, the factors have been arranged so that the sequence of first subscripts is the natural order 1, 2, ..., n ; the sequence j_1, j_2, \dots, j_n of second subscripts is then some one of the $n!$ permutations of the integers 1, 2, ..., n . (Facility will be gained if the reader will parallel the work of this section beginning with a product arranged so that the sequence of second subscripts is in natural order.)

For a given permutation j_1, j_2, \dots, j_n of the second subscripts, define $\epsilon_{j_1 j_2 \dots j_n} = +1$ or -1 according as the permutation is even or odd and form the signed product

$$(3.5) \quad \epsilon_{j_1 j_2 \dots j_n} a_{1j_1} a_{2j_2} \cdots a_{nj_n}$$

By the **determinant** of A , denoted by $|A|$, is meant the sum of all the different signed products of the form (3.5), called **terms** of $|A|$, which can be formed from the elements of A ; thus,

$$(3.6) \quad |A| = \sum_{\rho} \epsilon_{j_1 j_2 \dots j_n} a_{1j_1} a_{2j_2} \cdots a_{nj_n}$$

where the summation extends over $\rho = n!$ permutations $j_1 j_2 \dots j_n$ of the integers 1, 2, ..., n .

The determinant of a square matrix of order n is called a determinant of order n .

DETERMINANTS OF ORDER TWO AND THREE. From (3.6) we have for $n=2$ and $n=3$,

$$(3.7) \quad \begin{vmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{vmatrix} = \epsilon_{12} a_{11}a_{22} + \epsilon_{21} a_{12}a_{21} = a_{11}a_{22} - a_{12}a_{21}$$

and

$$(3.8) \quad \begin{vmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{vmatrix} = \epsilon_{123} a_{11}a_{22}a_{33} + \epsilon_{132} a_{11}a_{23}a_{32} + \epsilon_{213} a_{12}a_{21}a_{33} \\ + \epsilon_{231} a_{12}a_{23}a_{31} + \epsilon_{312} a_{13}a_{21}a_{32} + \epsilon_{321} a_{13}a_{22}a_{31} \\ = a_{11}a_{22}a_{33} - a_{11}a_{23}a_{32} - a_{12}a_{21}a_{33} \\ + a_{12}a_{23}a_{31} + a_{13}a_{21}a_{32} - a_{13}a_{22}a_{31} \\ = a_{11}(a_{22}a_{33} - a_{23}a_{32}) - a_{12}(a_{21}a_{33} - a_{23}a_{31}) + a_{13}(a_{21}a_{32} - a_{22}a_{31}) \\ = a_{11} \begin{vmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{vmatrix} - a_{12} \begin{vmatrix} a_{21} & a_{23} \\ a_{31} & a_{33} \end{vmatrix} + a_{13} \begin{vmatrix} a_{21} & a_{22} \\ a_{31} & a_{32} \end{vmatrix}$$

Example 1.

$$(a) \quad \begin{vmatrix} 1 & 2 \\ 3 & 4 \end{vmatrix} = 1 \cdot 4 - 2 \cdot 3 = 4 - 6 = -2$$

$$(b) \quad \begin{vmatrix} 2 & -1 \\ 3 & 0 \end{vmatrix} = 2 \cdot 0 - (-1) \cdot 3 = 0 + 3 = 3$$

$$(c) \quad \begin{vmatrix} 2 & 3 & 5 \\ 1 & 0 & 1 \\ 2 & 1 & 0 \end{vmatrix} = 2 \begin{vmatrix} 0 & 1 \\ 1 & 0 \end{vmatrix} - 3 \begin{vmatrix} 1 & 1 \\ 2 & 0 \end{vmatrix} + 5 \begin{vmatrix} 1 & 0 \\ 2 & 1 \end{vmatrix} \\ = 2(0 \cdot 0 - 1 \cdot 1) - 3(1 \cdot 0 - 1 \cdot 2) + 5(1 \cdot 1 - 0 \cdot 2) = 2(-1) - 3(-2) + 5(1) = 9$$

$$(d) \quad \begin{vmatrix} 2 & -3 & -4 \\ 1 & 0 & -2 \\ 0 & -5 & -6 \end{vmatrix} = 2\{0(-6) - (-2)(-5)\} - (-3)\{1(-6) - (-2)0\} + (-4)\{1(-5) - 0 \cdot 0\} \\ = -20 - 18 + 20 = -18$$

See Problem 1.

PROPERTIES OF DETERMINANTS. Throughout this section, A is the square matrix whose determinant $|A|$ is given by (3.6).

Suppose that every element of the i th row (every element of the j th column) is zero. Since every term of (3.6) contains one element from this row (column), every term in the sum is zero and we have

I. If every element of a row (column) of a square matrix A is zero, then $|A| = 0$.

Consider the transpose A' of A . It can be seen readily that every term of (3.6) can be obtained from A' by choosing properly the factors in order from the first, second, ... columns. Thus,

II. If A is a square matrix then $|A'| = |A|$; that is, for every theorem concerning the rows of a determinant there is a corresponding theorem concerning the columns and vice versa.

Denote by B the matrix obtained by multiplying each of the elements of the i th row of A by a scalar k . Since each term in the expansion of $|B|$ contains one and only one element from its i th row, that is, one and only one element having k as a factor,

$$|B| = k \sum_p \{\epsilon_{j_1 j_2 \dots j_n} a_{1 j_1} a_{2 j_2} \dots a_{n j_n}\} = k |A|$$

Thus,

III. If every element of a row (column) of a determinant $|A|$ is multiplied by a scalar k , the determinant is multiplied by k ; if every element of a row (column) of a determinant $|A|$ has k as a factor then k may be factored from $|A|$. For example,

$$\begin{vmatrix} a_{11} & ka_{12} & a_{13} \\ a_{21} & ka_{22} & a_{23} \\ a_{31} & ka_{32} & a_{33} \end{vmatrix} = k \begin{vmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{vmatrix} = \begin{vmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ ka_{31} & ka_{32} & ka_{33} \end{vmatrix}$$

Let B denote the matrix obtained from A by interchanging its i th and $(i+1)$ st rows. Each product in (3.6) of $|A|$ is a product of $|B|$, and vice versa; hence, except possibly for signs, (3.6) is the expansion of $|B|$. In counting the inversions in subscripts of any term of (3.6) as a term of $|B|$, i before $i+1$ in the row subscripts is an inversion; thus, each product of (3.6) with its sign changed is a term of $|B|$ and $|B| = -|A|$. Hence,

IV. If B is obtained from A by interchanging any two adjacent rows (columns), then $|B| = -|A|$.

As a consequence of Theorem IV, we have

V. If B is obtained from A by interchanging any two of its rows (columns), then $|B| = -|A|$.

VI. If B is obtained from A by carrying its i th row (column) over p rows (columns), then $|B| = (-1)^p |A|$.

VII. If two rows (columns) of A are identical, then $|A| = 0$.

Suppose that each element of the first row of A is expressed as a binomial $a_{1j} = b_{1j} + c_{1j}$, ($j = 1, 2, \dots, n$). Then

$$\begin{aligned} |A| &= \sum_p \epsilon_{j_1 j_2 \dots j_n} (b_{1j_1} + c_{1j_1}) a_{2j_2} a_{3j_3} \dots a_{nj_n} \\ &= \sum_p \epsilon_{j_1 j_2 \dots j_n} b_{1j_1} a_{2j_2} a_{3j_3} \dots a_{nj_n} + \sum_p \epsilon_{j_1 j_2 \dots j_n} c_{1j_1} a_{2j_2} a_{3j_3} \dots a_{nj_n} \\ &= \begin{vmatrix} b_{11} & b_{12} & b_{13} & \dots & b_{1n} \\ a_{21} & a_{22} & a_{23} & \dots & a_{2n} \\ \dots & \dots & \dots & \dots & \dots \\ a_{n1} & a_{n2} & a_{n3} & \dots & a_{nn} \end{vmatrix} + \begin{vmatrix} c_{11} & c_{12} & c_{13} & \dots & c_{1n} \\ a_{21} & a_{22} & a_{23} & \dots & a_{2n} \\ \dots & \dots & \dots & \dots & \dots \\ a_{n1} & a_{n2} & a_{n3} & \dots & a_{nn} \end{vmatrix} \end{aligned}$$

In general,

VIII. If every element of the i th row (column) of A is the sum of p terms, then $|A|$ can be expressed as the sum of p determinants. The elements in the i th rows (columns) of these p determinants are respectively the first, second, ..., p th terms of the sums and all other rows (columns) are those of A .

The most useful theorem is

IX. If B is obtained from A by adding to the elements of its i th row (column), a scalar multiple of the corresponding elements of another row (column), then $|B| = |A|$. For example,

$$\begin{vmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{vmatrix} = \begin{vmatrix} a_{11} + ka_{13} & a_{12} & a_{13} \\ a_{21} + ka_{23} & a_{22} & a_{23} \\ a_{31} + ka_{33} & a_{32} & a_{33} \end{vmatrix} = \begin{vmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} + ka_{21} & a_{32} + ka_{22} & a_{33} + ka_{23} \end{vmatrix}$$

See Problems 2-7.

FIRST MINORS AND COFACTORS. Let A be the n -square matrix (3.3) whose determinant $|A|$ is given by (3.6). When from A the elements of its i th row and j th column are removed, the determinant of the remaining $(n-1)$ -square matrix is called a **first minor** of A or of $|A|$ and denoted by $|M_{ij}|$.

More frequently, it is called the minor of a_{ij} . The signed minor, $(-1)^{i+j} |M_{ij}|$ is called the cofactor of a_{ij} and is denoted by α_{ij} .

Example 2. If $A = \begin{vmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{vmatrix}$,

$$|M_{11}| = \begin{vmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{vmatrix}, \quad |M_{12}| = \begin{vmatrix} a_{21} & a_{23} \\ a_{31} & a_{33} \end{vmatrix}, \quad |M_{13}| = \begin{vmatrix} a_{21} & a_{22} \\ a_{31} & a_{32} \end{vmatrix}$$

and

$$\alpha_{11} = (-1)^{1+1} |M_{11}| = |M_{11}|, \quad \alpha_{12} = (-1)^{1+2} |M_{12}| = -|M_{12}|,$$

$$\alpha_{13} = (-1)^{1+3} |M_{13}| = |M_{13}|$$

Then (3.8) is

$$|A| = a_{11}|M_{11}| - a_{12}|M_{12}| + a_{13}|M_{13}|$$

$$= a_{11}\alpha_{11} + a_{12}\alpha_{12} + a_{13}\alpha_{13}$$

In Problem 9, we prove

X. The value of the determinant $|A|$, where A is the matrix of (3.3), is the sum of the products obtained by multiplying each element of a row (column) of $|A|$ by its cofactor, i.e.,

$$(3.9) \quad |A| = a_{i1}\alpha_{i1} + a_{i2}\alpha_{i2} + \dots + a_{in}\alpha_{in} = \sum_{k=1}^n a_{ik}\alpha_{ik}$$

$$(3.10) \quad |A| = a_{1j}\alpha_{1j} + a_{2j}\alpha_{2j} + \dots + a_{nj}\alpha_{nj} = \sum_{k=1}^n a_{kj}\alpha_{kj} \quad (i, j, = 1, 2, \dots, n)$$

Using Theorem VII, we can prove

XI. The sum of the products formed by multiplying the elements of a row (column) of an n -square matrix A by the corresponding cofactors of another row (column) of A is zero.

Example 3. If A is the matrix of Example 2, we have

$$\begin{aligned} \text{and} \quad & a_{31}\alpha_{31} + a_{32}\alpha_{32} + a_{33}\alpha_{33} = |A| \\ \text{while} \quad & a_{12}\alpha_{12} + a_{22}\alpha_{22} + a_{32}\alpha_{32} = |A| \\ \text{and} \quad & a_{31}\alpha_{21} + a_{32}\alpha_{22} + a_{33}\alpha_{23} = 0 \\ & a_{12}\alpha_{13} + a_{22}\alpha_{23} + a_{32}\alpha_{33} = 0 \end{aligned}$$

See Problems 10-11.

MINORS AND ALGEBRAIC COMPLEMENTS. Consider the matrix (3.3). Let i_1, i_2, \dots, i_m , arranged in order of magnitude, be m , ($1 \leq m < n$), of the row indices $1, 2, \dots, n$ and let j_1, j_2, \dots, j_m arranged in order of magnitude, be m of the column indices. Let the remaining row and column indices, arranged in order of magnitude, be respectively $i_{m+1}, i_{m+2}, \dots, i_n$ and $j_{m+1}, j_{m+2}, \dots, j_n$. Such a separation of the row and column indices determines uniquely two matrices

$$(3.11) \quad A_{\substack{j_1, j_2, \dots, j_m \\ i_1, i_2, \dots, i_m}} = \begin{bmatrix} a_{i_1, j_1} & a_{i_1, j_2} & \dots & a_{i_1, j_m} \\ a_{i_2, j_1} & a_{i_2, j_2} & \dots & a_{i_2, j_m} \\ \dots & \dots & \dots & \dots \\ a_{i_m, j_1} & a_{i_m, j_2} & \dots & a_{i_m, j_m} \end{bmatrix}$$

and

$$(3.12) \quad A_{i_{m+1}, i_{m+2}, \dots, i_n}^{j_{m+1}, j_{m+2}, \dots, j_n} = \begin{bmatrix} a_{i_{m+1}, j_{m+1}} & a_{i_{m+1}, j_{m+2}} & \dots & a_{i_{m+1}, j_n} \\ a_{i_{m+2}, j_{m+1}} & a_{i_{m+2}, j_{m+2}} & \dots & a_{i_{m+2}, j_n} \\ \dots & \dots & \dots & \dots \\ a_{i_n, j_{m+1}} & a_{i_n, j_{m+2}} & \dots & a_{i_n, j_n} \end{bmatrix}$$

called sub-matrices of A .

The determinant of each of these sub-matrices is called a minor of A and the pair of minors

$$\begin{vmatrix} j_1, j_2, \dots, j_m \\ A_{i_1, i_2, \dots, i_m} \end{vmatrix} \quad \text{and} \quad \begin{vmatrix} j_{m+1}, j_{m+2}, \dots, j_n \\ A_{i_{m+1}, i_{m+2}, \dots, i_n} \end{vmatrix}$$

are called **complementary minors** of A , each being the complement of the other.

Example 3. For the 5-square matrix $A = [a_{ij}]$,

$$\begin{vmatrix} 1, 3 \\ A_{2, 5} \end{vmatrix} = \begin{vmatrix} a_{21} & a_{23} \\ a_{51} & a_{53} \end{vmatrix} \quad \text{and} \quad \begin{vmatrix} 2, 4, 5 \\ A_{1, 3, 4} \end{vmatrix} = \begin{vmatrix} a_{12} & a_{14} & a_{15} \\ a_{32} & a_{34} & a_{35} \\ a_{42} & a_{44} & a_{45} \end{vmatrix}$$

are a pair of complementary minors.

Let

$$(3.13) \quad p = i_1 + i_2 + \dots + i_m + j_1 + j_2 + \dots + j_m$$

and

$$(3.14) \quad q = i_{m+1} + i_{m+2} + \dots + i_n + j_{m+1} + j_{m+2} + \dots + j_n$$

The signed minor $(-1)^p \begin{vmatrix} j_1, j_2, \dots, j_m \\ A_{i_1, i_2, \dots, i_m} \end{vmatrix}$ is called the **algebraic complement** of

and $(-1)^q \begin{vmatrix} j_{m+1}, j_{m+2}, \dots, j_n \\ A_{i_{m+1}, i_{m+2}, \dots, i_n} \end{vmatrix}$ is called the **algebraic complement** of

$$\begin{vmatrix} j_1, j_2, \dots, j_m \\ A_{i_1, i_2, \dots, i_m} \end{vmatrix}$$

Example 4. For the minors of Example 3, $(-1)^{2+5+1+3} |A_{2,5}^{1,3}| = -|A_{2,5}^{1,3}|$ is the algebraic complement of $|A_{1,3,4}^{2,4,5}|$ and $(-1)^{1+3+4+2+4+5} |A_{1,3,4}^{2,4,5}| = -|A_{1,3,4}^{2,4,5}|$ is the algebraic complement of $|A_{2,5}^{1,3}|$. Note that the sign given to the two complementary minors is the same. Is this always true?

When $m=1$, (3.11) becomes $A_{i_1}^{j_1} = [a_{i_1 j_1}]$ and $|A_{i_1}^{j_1}| = a_{i_1 j_1}$, an element of A . The complementary minor $\begin{vmatrix} j_2, j_3, \dots, j_n \\ A_{i_2, i_3, \dots, i_n} \end{vmatrix}$ is $|M_{i_1, j_1}|$ in the notation of the section above, and the algebraic complement is the cofactor $\alpha_{i_1 j_1}$.

A minor of A , whose diagonal elements are also diagonal elements of A , is called a **principal minor** of A . The complement of a principal minor of A is also a principal minor of A ; the algebraic complement of a principal minor is its complement.

Example 5. For the 5-square matrix $A = [a_{ij}]$,

$$|A_{1,3}^{1,3}| = \begin{vmatrix} a_{11} & a_{13} \\ a_{31} & a_{33} \end{vmatrix} \quad \text{and} \quad |A_{2,4,5}^{2,4,5}| = \begin{vmatrix} a_{22} & a_{24} & a_{25} \\ a_{42} & a_{44} & a_{45} \\ a_{52} & a_{54} & a_{55} \end{vmatrix}$$

are a pair of complementary principal minors of A . What is the algebraic complement of each?

The terms minor, complementary minor, algebraic complement, and principal minor as defined above for a square matrix A will also be used without change in connection with $|A|$.

See Problems 12-13.

SOLVED PROBLEMS

1. (a) $\begin{vmatrix} 2 & 3 \\ -1 & 4 \end{vmatrix} = 2 \cdot 4 - 3(-1) = 11$

(b) $\begin{vmatrix} 1 & 0 & 2 \\ 3 & 4 & 5 \\ 5 & 6 & 7 \end{vmatrix} = (1) \begin{vmatrix} 4 & 5 \\ 6 & 7 \end{vmatrix} - 0 \begin{vmatrix} 3 & 5 \\ 5 & 7 \end{vmatrix} + 2 \begin{vmatrix} 3 & 4 \\ 5 & 6 \end{vmatrix} = (1)(4 \cdot 7 - 5 \cdot 6) - 0 + 2(3 \cdot 6 - 4 \cdot 5) = -2 - 4 = -6$

(c) $\begin{vmatrix} 1 & 0 & 6 \\ 3 & 4 & 15 \\ 5 & 6 & 21 \end{vmatrix} = 1(4 \cdot 21 - 15 \cdot 6) + 6(3 \cdot 6 - 4 \cdot 5) = -18$

(d) $\begin{vmatrix} 1 & 0 & 0 \\ 2 & 3 & 5 \\ 4 & 1 & 3 \end{vmatrix} = 1(3 \cdot 3 - 5 \cdot 1) = 4$

2. Adding to the elements of the first column the corresponding elements of the other columns,

$$\begin{vmatrix} -4 & 1 & 1 & 1 & 1 \\ 1 & -4 & 1 & 1 & 1 \\ 1 & 1 & -4 & 1 & 1 \\ 1 & 1 & 1 & -4 & 1 \\ 1 & 1 & 1 & 1 & -4 \end{vmatrix} = \begin{vmatrix} 0 & 1 & 1 & 1 & 1 \\ 0 & -4 & 1 & 1 & 1 \\ 0 & 1 & -4 & 1 & 1 \\ 0 & 1 & 1 & -4 & 1 \\ 0 & 1 & 1 & 1 & -4 \end{vmatrix} = 0$$

by Theorem I.

3. Adding the second column to the third, removing the common factor from this third column, and using Theorem VII

$$\begin{vmatrix} 1 & a & b+c \\ 1 & b & c+a \\ 1 & c & a+b \end{vmatrix} = \begin{vmatrix} 1 & a & a+b+c \\ 1 & b & a+b+c \\ 1 & c & a+b+c \end{vmatrix} = (a+b+c) \begin{vmatrix} 1 & a & 1 \\ 1 & b & 1 \\ 1 & c & 1 \end{vmatrix} = 0$$

4. Adding to the third row the first and second rows, then removing the common factor 2; subtracting the second row from the third; subtracting the third row from the first; subtracting the first row from the second; finally carrying the third row over the other rows

$$\begin{vmatrix} a_1+b_1 & a_2+b_2 & a_3+b_3 \\ b_1+c_1 & b_2+c_2 & b_3+c_3 \\ c_1+a_1 & c_2+a_2 & c_3+a_3 \end{vmatrix} = 2 \begin{vmatrix} a_1+b_1 & a_2+b_2 & a_3+b_3 \\ b_1+c_1 & b_2+c_2 & b_3+c_3 \\ a_1+b_1+c_1 & a_2+b_2+c_2 & a_3+b_3+c_3 \end{vmatrix} = 2 \begin{vmatrix} a_1+b_1 & a_2+b_2 & a_3+b_3 \\ b_1+c_1 & b_2+c_2 & b_3+c_3 \\ a_1 & a_2 & a_3 \end{vmatrix}$$

$$= 2 \begin{vmatrix} b_1 & b_2 & b_3 \\ b_1+c_1 & b_2+c_2 & b_3+c_3 \\ a_1 & a_2 & a_3 \end{vmatrix} = 2 \begin{vmatrix} b_1 & b_2 & b_3 \\ c_1 & c_2 & c_3 \\ a_1 & a_2 & a_3 \end{vmatrix} = 2 \begin{vmatrix} a_1 & a_2 & a_3 \\ b_1 & b_2 & b_3 \\ c_1 & c_2 & c_3 \end{vmatrix}$$

5. Without expanding, show that $|A| = \begin{vmatrix} a_1^2 & a_1 & 1 \\ a_2^2 & a_2 & 1 \\ a_3^2 & a_3 & 1 \end{vmatrix} = -(a_1 - a_2)(a_2 - a_3)(a_3 - a_1)$.

Subtract the second row from the first; then

$$|A| = \begin{vmatrix} a_1^2 - a_2^2 & a_1 - a_2 & 0 \\ a_2^2 & a_2 & 1 \\ a_3^2 & a_3 & 1 \end{vmatrix} = (a_1 - a_2) \begin{vmatrix} a_1 + a_2 & 1 & 0 \\ a_2^2 & a_2 & 1 \\ a_3^2 & a_3 & 1 \end{vmatrix} \quad \text{by Theorem III}$$

and $a_1 - a_2$ is a factor of $|A|$. Similarly, $a_2 - a_3$ and $a_3 - a_1$ are factors. Now $|A|$ is of order three in the letters; hence,

$$(i) \quad |A| = k(a_1 - a_2)(a_2 - a_3)(a_3 - a_1)$$

The product of the diagonal elements, $a_1^2 a_2$, is a term of $|A|$ and, from (i), the term is $-k a_1^2 a_2$. Thus, $k = -1$ and $|A| = -(a_1 - a_2)(a_2 - a_3)(a_3 - a_1)$. Note that $|A|$ vanishes if and only if two of the a_1, a_2, a_3 are equal.

6. Prove: If A is skew-symmetric and of odd order $2p - 1$, then $|A| = 0$.

Since A is skew-symmetric, $A' = -A$; then $|A'| = |-A| = (-1)^{2p-1} |A| = -|A|$. But, by Theorem II, $|A'| = |A|$; hence, $|A| = -|A|$ and $|A| = 0$.

7. Prove: If A is Hermitian, then $|A|$ is a real number.

Since A is Hermitian, $\bar{A} = A'$, and $|\bar{A}| = |A'| = |A|$ by Theorem II. But if

$$|A| = \sum_p \epsilon_{j_1 j_2 \dots j_n} a_{1j_1} a_{2j_2} \dots a_{nj_n} = a + bi$$

then

$$|\bar{A}| = \sum_p \epsilon_{j_1 j_2 \dots j_n} \bar{a}_{1j_1} \bar{a}_{2j_2} \dots \bar{a}_{nj_n} = a - bi$$

Now $|\bar{A}| = |A|$ requires $b = 0$; hence, $|A|$ is a real number.

8. For the matrix $A = \begin{bmatrix} 1 & 2 & 3 \\ 2 & 3 & 2 \\ 1 & 2 & 2 \end{bmatrix}$,

$$\alpha_{11} = (-1)^{1+1} \begin{vmatrix} 3 & 2 \\ 2 & 2 \end{vmatrix} = 2, \quad \alpha_{12} = (-1)^{1+2} \begin{vmatrix} 2 & 2 \\ 1 & 2 \end{vmatrix} = -2, \quad \alpha_{13} = (-1)^{1+3} \begin{vmatrix} 2 & 3 \\ 1 & 2 \end{vmatrix} = 1$$

$$\alpha_{21} = (-1)^{2+1} \begin{vmatrix} 2 & 3 \\ 1 & 2 \end{vmatrix} = 2, \quad \alpha_{22} = (-1)^{2+2} \begin{vmatrix} 1 & 3 \\ 1 & 2 \end{vmatrix} = -1, \quad \alpha_{23} = (-1)^{2+3} \begin{vmatrix} 1 & 2 \\ 1 & 2 \end{vmatrix} = 0$$

$$\alpha_{31} = (-1)^{3+1} \begin{vmatrix} 2 & 3 \\ 3 & 2 \end{vmatrix} = -5, \quad \alpha_{32} = (-1)^{3+2} \begin{vmatrix} 1 & 3 \\ 2 & 2 \end{vmatrix} = 4, \quad \alpha_{33} = (-1)^{3+3} \begin{vmatrix} 1 & 2 \\ 2 & 3 \end{vmatrix} = -1$$

Note that the signs given to the minors of the elements in forming the cofactors follow the pattern

$$\begin{matrix} + & - & + \\ - & + & - \\ + & - & + \end{matrix}$$

where each sign occupies the same position in the display as the element, whose cofactor is required, occupies in A . Write the display of signs for a 5-square matrix.

9. Prove: The value of the determinant $|A|$ of an n -square matrix A is the sum of the products obtained by multiplying each element of a row (column) of A by its cofactor.

We shall prove this for a row. The terms of (3.6) having a_{11} as a factor are

(a)
$$a_{11} \sum \epsilon_{1, j_2 j_3 \dots j_n} a_{2j_2} a_{3j_3} \dots a_{nj_n}$$

Now $\epsilon_{1, j_2 j_3 \dots j_n} = \epsilon_{j_2 j_3 \dots j_n}$ since in a permutation $1, j_1, j_2, \dots, j_n$, the 1 is in natural order. Then (a) may be written as

(b)
$$a_{11} \sum_{\sigma} \epsilon_{j_2 j_3 \dots j_n} a_{2j_2} a_{3j_3} \dots a_{nj_n}$$

where the summation extends over the $\sigma = (n-1)!$ permutations of the integers $2, 3, \dots, n$, and hence, as

(c)
$$a_{11} \begin{vmatrix} a_{22} & a_{23} & \dots & a_{2n} \\ a_{32} & a_{33} & \dots & a_{3n} \\ \dots & \dots & \dots & \dots \\ a_{n2} & a_{n3} & \dots & a_{nn} \end{vmatrix} = a_{11} |M_{11}|$$

Consider the matrix B obtained from A by moving its s th column over the first $s-1$ columns. By Theorem VI, $|B| = (-1)^{s-1} |A|$. Moreover, the element standing in the first row and first column of B is a_{1s} and the minor of a_{1s} in B is precisely the minor $|M_{1s}|$ of a_{1s} in A . By the argument leading to (c), the terms of $a_{1s} |M_{1s}|$ are all the terms of $|B|$ having a_{1s} as a factor and, thus, all the terms of $(-1)^{s-1} |A|$ having a_{1s} as a factor. Then the terms of $a_{1s} \{(-1)^{s-1} |M_{1s}|\}$ are all the terms of $|A|$ having a_{1s} as a factor. Thus,

(3.15)
$$\begin{aligned} |A| &= a_{11} \{(-1)^{1+1} |M_{11}|\} + a_{12} \{(-1)^{1+2} |M_{12}|\} \\ &\quad + \dots + a_{1s} \{(-1)^{1+s} |M_{1s}|\} + \dots + a_{1n} \{(-1)^{1+n} |M_{1n}|\} \\ &= a_{11} \alpha_{11} + a_{12} \alpha_{12} + \dots + a_{1n} \alpha_{1n} \end{aligned}$$

since $(-1)^{s-1} = (-1)^{s+1}$. We have (3.9) with $i = 1$. We shall call (3.15) the expansion of $|A|$ along its first row.

The expansion of $|A|$ along its r th row (that is, (3.9) for $i=r$) is obtained by repeating the above arguments. Let B be the matrix obtained from A by moving its r th row over the first $r-1$ rows and then its s th column over the first $s-1$ columns. Then

$$|B| = (-1)^{r-1} \cdot (-1)^{s-1} |A| = (-1)^{r+s} |A|$$

The element standing in the first row and the first column of B is a_{rs} and the minor of a_{rs} in B is precisely the minor of a_{rs} in A . Thus, the terms of

$$a_{rs} \{(-1)^{r+s} |M_{rs}|\}$$

are all the terms of $|A|$ having a_{rs} as a factor. Then

$$|A| = \sum_{k=1}^n a_{rk} \{(-1)^{r+k} |M_{rk}|\} = \sum_{k=1}^n a_{rk} \alpha_{rk}$$

and we have (3.9) for $i=r$.

10. When α_{ij} is the cofactor of a_{ij} in the n -square matrix $A = [a_{ij}]$, show that

$$(i) \quad k_1 \alpha_{1j} + k_2 \alpha_{2j} + \dots + k_n \alpha_{nj} = \begin{vmatrix} a_{11} & a_{12} & \dots & a_{1,j-1} & k_1 & a_{1,j+1} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2,j-1} & k_2 & a_{2,j+1} & \dots & a_{2n} \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ a_{n1} & a_{n2} & \dots & a_{n,j-1} & k_n & a_{n,j+1} & \dots & a_{nn} \end{vmatrix}$$

This relation follows from (3.10) by replacing a_{1j} with k_1 , a_{2j} with k_2 , ..., a_{nj} with k_n . In making these replacements none of the cofactors $\alpha_{1j}, \alpha_{2j}, \dots, \alpha_{nj}$ appearing is affected since none contains an element from the j th column of A .

By Theorem VII, the determinant in (i) is 0 when $k_r = a_{rs}$, ($r = 1, 2, \dots, n$ and $s \neq j$). By Theorems VIII and VII, the determinant in (i) is $|A|$ when $k_r = a_{rj} + ka_{rs}$, ($r = 1, 2, \dots, n$ and $s \neq j$).

Write the equality similar to (i) obtained from (3.9) when the elements of the i th row of A are replaced by k_1, k_2, \dots, k_n .

$$11. \text{ Evaluate: } (a) \quad |A| = \begin{vmatrix} 1 & 0 & 2 \\ 3 & 0 & 4 \\ 2 & -5 & 1 \end{vmatrix} \quad (c) \quad |A| = \begin{vmatrix} 3 & 4 & 5 \\ 1 & 2 & 3 \\ -2 & 5 & -4 \end{vmatrix} \quad (e) \quad |A| = \begin{vmatrix} 28 & 25 & 38 \\ 42 & 38 & 65 \\ 56 & 47 & 83 \end{vmatrix}$$

$$(b) \quad |A| = \begin{vmatrix} 1 & 4 & 3 \\ -2 & 1 & 5 \\ -3 & 2 & 4 \end{vmatrix} \quad (d) \quad |A| = \begin{vmatrix} 2 & 3 & -4 \\ 5 & -6 & 3 \\ 4 & 2 & -3 \end{vmatrix}$$

(a) Expanding along the second column (see Theorem X)

$$\begin{aligned} |A| &= \begin{vmatrix} 1 & 0 & 2 \\ 3 & 0 & 4 \\ 2 & -5 & 1 \end{vmatrix} = a_{12}\alpha_{12} + a_{22}\alpha_{22} + a_{32}\alpha_{32} = 0 \cdot \alpha_{12} + 0 \cdot \alpha_{22} + (-5)\alpha_{32} \\ &= -5(-1)^{3+2} \begin{vmatrix} 1 & 2 \\ 3 & 4 \end{vmatrix} = 5(4-6) = -10 \end{aligned}$$

(b) Subtracting twice the second column from the third (see Theorem IX)

$$\begin{aligned} |A| &= \begin{vmatrix} 1 & 4 & 3 \\ -2 & 1 & 5 \\ -3 & 2 & 4 \end{vmatrix} = \begin{vmatrix} 1 & 4 & 3-2 \cdot 4 \\ -2 & 1 & 5-2 \cdot 1 \\ -3 & 2 & 4-2 \cdot 2 \end{vmatrix} = \begin{vmatrix} 1 & 4 & 0 \\ -2 & 1 & 3 \\ -3 & 2 & 0 \end{vmatrix} = 3(-1)^{2+3} \begin{vmatrix} 1 & 4 \\ -3 & 2 \end{vmatrix} \\ &= -3(14) = -42 \end{aligned}$$

(c) Subtracting three times the second row from the first and adding twice the second row to the third

$$\begin{aligned} |A| &= \begin{vmatrix} 3 & 4 & 5 \\ 1 & 2 & 3 \\ -2 & 5 & -4 \end{vmatrix} = \begin{vmatrix} 3-3(1) & 4-3(2) & 5-3(3) \\ 1 & 2 & 3 \\ -2+2(1) & 5+2(2) & -4+2(3) \end{vmatrix} = \begin{vmatrix} 0 & -2 & -4 \\ 1 & 2 & 3 \\ 0 & 9 & 2 \end{vmatrix} = - \begin{vmatrix} -2 & -4 \\ 9 & 2 \end{vmatrix} \\ &= -(-4+36) = -32 \end{aligned}$$

(d) Subtracting the first column from the second and then proceeding as in (c)

$$\begin{aligned} |A| &= \begin{vmatrix} 2 & 3 & -4 \\ 5 & -6 & 3 \\ 4 & 2 & -3 \end{vmatrix} = \begin{vmatrix} 2 & 3-1 & -4 \\ 5 & -6-1 & 3 \\ 4 & 2-1 & -3 \end{vmatrix} = \begin{vmatrix} 2-2(1) & 1 & -4+4(1) \\ 5-2(-11) & -11 & 3+4(-11) \\ 4-2(-2) & -2 & -3+4(-2) \end{vmatrix} \\ &= \begin{vmatrix} 0 & 1 & 0 \\ 27 & -11 & -41 \\ 8 & -2 & -11 \end{vmatrix} = - \begin{vmatrix} 27 & -41 \\ 8 & -11 \end{vmatrix} = -31 \end{aligned}$$

(e) Factoring 14 from the first column, then using Theorem IX to reduce the elements in the remaining columns

$$\begin{aligned}
 |A| &= \begin{vmatrix} 28 & 25 & 38 \\ 42 & 38 & 65 \\ 56 & 47 & 83 \end{vmatrix} = 14 \begin{vmatrix} 2 & 25 & 38 \\ 3 & 38 & 65 \\ 4 & 47 & 83 \end{vmatrix} = 14 \begin{vmatrix} 2 & 25-12(2) & 38-20(2) \\ 3 & 38-12(3) & 65-20(3) \\ 4 & 47-12(4) & 83-20(4) \end{vmatrix} \\
 &= 14 \begin{vmatrix} 2 & 1 & -2 \\ 3 & 2 & 5 \\ 4 & -1 & 3 \end{vmatrix} = 14 \begin{vmatrix} 0 & 1 & 0 \\ -1 & 2 & 9 \\ 6 & -1 & 1 \end{vmatrix} = -14 \begin{vmatrix} -1 & 9 \\ 6 & 1 \end{vmatrix} = -14(-1-54) = 770
 \end{aligned}$$

12. Show that p and q , given by (3.13) and (3.14), are either both even or both odd.

Since each row (column) index is found in either p or q but never in both,

$$p + q = (1+2+\dots+n) + (1+2+\dots+n) = 2 \cdot \frac{1}{2}n(n+1) = n(n+1)$$

Now $p+q$ is even (either n or $n+1$ is even); hence, p and q are either both even or both odd. Thus, $(-1)^p = (-1)^q$ and only one need be computed.

13. For the matrix $A = [a_{ij}] = \begin{bmatrix} 1 & 2 & 3 & 4 & 5 \\ 6 & 7 & 8 & 9 & 10 \\ 11 & 12 & 13 & 14 & 15 \\ 16 & 17 & 18 & 19 & 20 \\ 21 & 22 & 23 & 24 & 25 \end{bmatrix}$, the algebraic complement of $|A_{2,3}^{2,4}|$ is

$$(-1)^{2+3+2+4} |A_{1,4,5}^{1,3,5}| = - \begin{vmatrix} 1 & 3 & 5 \\ 16 & 18 & 20 \\ 21 & 23 & 25 \end{vmatrix} \quad (\text{see Problem 12})$$

and the algebraic complement of $|A_{1,4,5}^{1,3,5}|$ is $-|A_{2,3}^{2,4}| = - \begin{vmatrix} 7 & 9 \\ 12 & 14 \end{vmatrix}$.

SUPPLEMENTARY PROBLEMS

14. Show that the permutation 12534 of the integers 1, 2, 3, 4, 5 is even, 24135 is odd, 41532 is even, 53142 is odd, and 52314 is even.
15. List the complete set of permutations of 1, 2, 3, 4, taken together; show that half are even and half are odd.
16. Let the elements of the diagonal of a 5-square matrix A be a, b, c, d, e . Show, using (3.6), that when A is diagonal, upper triangular, or lower triangular then $|A| = abcde$.
17. Given $A = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$ and $B = \begin{bmatrix} 5 & 6 \\ 7 & 8 \end{bmatrix}$, show that $AB \neq BA \neq A'B \neq AB' \neq A'B' \neq B'A'$ but that the determinant of each product is 4.
18. Evaluate, as in Problem 1,

(a) $\begin{vmatrix} 2 & -1 & 1 \\ 3 & 2 & 4 \\ -1 & 0 & 3 \end{vmatrix} = 27$

(b) $\begin{vmatrix} 2 & 2 & -2 \\ 1 & 2 & 3 \\ 2 & 3 & 4 \end{vmatrix} = 4$

(c) $\begin{vmatrix} 0 & 2 & 3 \\ -2 & 0 & 4 \\ -3 & -4 & 0 \end{vmatrix} = 0$

19. (a) Evaluate $|A| = \begin{vmatrix} 1 & 2 & 10 \\ 2 & 3 & 9 \\ 4 & 5 & 11 \end{vmatrix} = -4$.
- (b) Denote by $|B|$ the determinant obtained from $|A|$ by multiplying the elements of its second column by 5. Evaluate $|B|$ to verify Theorem III.
- (c) Denote by $|C|$ the determinant obtained from $|A|$ by interchanging its first and third columns. Evaluate $|C|$ to verify Theorem V.
- (d) Show that $|A| = \begin{vmatrix} 1 & 2 & 7 \\ 2 & 3 & 5 \\ 4 & 5 & 8 \end{vmatrix} + \begin{vmatrix} 1 & 2 & 3 \\ 2 & 3 & 4 \\ 4 & 5 & 3 \end{vmatrix}$, thus verifying Theorem VIII.
- (e) Obtain from $|A|$ the determinant $|D| = \begin{vmatrix} 1 & 2 & 7 \\ 2 & 3 & 3 \\ 4 & 5 & -1 \end{vmatrix}$ by subtracting three times the elements of the first column from the corresponding elements of the third column. Evaluate $|D|$ to verify Theorem IX.
- (f) In $|A|$ subtract twice the first row from the second and four times the first row from the third. Evaluate the resulting determinant.
- (g) In $|A|$ multiply the first column by three and from it subtract the third column. Evaluate to show that $|A|$ has been tripled. Compare with (e). Do not confuse (e) and (g).
20. If A is an n -square matrix and k is a scalar, use (3.6) to show that $|kA| = k^n |A|$.
21. Prove: (a) If $|A| = k$, then $|\overline{A}| = \overline{k} = |\overline{A}|$.
(b) If A is skew-Hermitian, then $|A|$ is either real or is a pure imaginary number.
22. (a) Count the number of interchanges of adjacent rows (columns) necessary to obtain B from A in Theorem V and thus prove the theorem.
(b) Same, for Theorem VI.
23. Prove Theorem VII. Hint: Interchange the identical rows and use Theorem V.
24. Prove: If any two rows (columns) of a square matrix A are proportional, then $|A| = 0$.
25. Use Theorems VIII, III, and VII to prove Theorem IX.
26. Evaluate the determinants of Problem 18 as in Problem 11.
27. Use (3.6) to evaluate $|A| = \begin{vmatrix} a & b & 0 & 0 \\ c & d & 0 & 0 \\ 0 & 0 & e & f \\ 0 & 0 & g & h \end{vmatrix}$; then check that $|A| = \begin{vmatrix} a & b \\ c & d \end{vmatrix} \begin{vmatrix} e & f \\ g & h \end{vmatrix}$. Thus, if $A = \text{diag}(A_1, A_2)$, where A_1, A_2 are 2-square matrices, $|A| = |A_1| \cdot |A_2|$.
28. Show that the cofactor of each element of $\begin{bmatrix} -1/3 & -2/3 & -2/3 \\ 2/3 & 1/3 & -2/3 \\ 2/3 & -2/3 & 1/3 \end{bmatrix}$ is that element.
29. Show that the cofactor of an element of any row of $\begin{bmatrix} -4 & -3 & -3 \\ 1 & 0 & 1 \\ 4 & 4 & 3 \end{bmatrix}$ is the corresponding element of the same numbered column.
30. Prove: (a) If A is symmetric then $a_{ij} = a_{ji}$ when $i \neq j$.
(b) If A is n -square and skew-symmetric then $a_{ij} = (-1)^{n-1} a_{ji}$ when $i \neq j$.

31. For the matrix A of Problem 8;

(a) show that $|A| = 1$

(b) form the matrix $C = \begin{bmatrix} \alpha_{11} & \alpha_{21} & \alpha_{31} \\ \alpha_{12} & \alpha_{22} & \alpha_{32} \\ \alpha_{13} & \alpha_{23} & \alpha_{33} \end{bmatrix}$ and show that $AC = I$.

(c) explain why the result in (b) is known as soon as (a) is known.

32. Multiply the columns of $|A| = \begin{vmatrix} bc & a^2 & a^2 \\ b^2 & ca & b^2 \\ c^2 & c^2 & ab \end{vmatrix}$ respectively by a, b, c ; remove the common factor from each of

the rows to show that $|A| = \begin{vmatrix} bc & ab & ca \\ ab & ca & bc \\ ca & bc & ab \end{vmatrix}$.

33. Without evaluating show that $\begin{vmatrix} a^2 & a & 1 & bcd \\ b^2 & b & 1 & acd \\ c^2 & c & 1 & abd \\ d^2 & d & 1 & abc \end{vmatrix} = \begin{vmatrix} a^3 & a^2 & a & 1 \\ b^3 & b^2 & b & 1 \\ c^3 & c^2 & c & 1 \\ d^3 & d^2 & d & 1 \end{vmatrix} = (a-b)(a-c)(a-d)(b-c)(b-d)(c-d)$.

34. Show that the n -square determinant $|A| = \begin{vmatrix} 0 & 1 & 1 & \dots & 1 \\ 1 & 0 & 1 & \dots & 1 \\ 1 & 1 & 0 & \dots & 1 \\ \dots & \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots & \dots \\ 1 & 1 & 1 & \dots & 0 \end{vmatrix} = (n-1) \begin{vmatrix} 0 & 1 & 1 & \dots & 1 & 1 \\ 1 & 0 & 1 & \dots & 1 & 1 \\ 1 & 1 & 0 & \dots & 1 & 1 \\ \dots & \dots & \dots & \dots & \dots & \dots \\ 1 & 1 & 1 & \dots & 0 & 1 \\ 1 & 1 & 1 & \dots & 1 & 1 \end{vmatrix} = (-1)^{n-1} (n-1)$.

35. Prove: $\begin{vmatrix} a_1^{n-1} & a_1^{n-2} & \dots & a_1 & 1 \\ a_2^{n-1} & a_2^{n-2} & \dots & a_2 & 1 \\ \dots & \dots & \dots & \dots & \dots \\ a_n^{n-1} & a_n^{n-2} & \dots & a_n & 1 \end{vmatrix} = \{(a_1 - a_2)(a_1 - a_3) \dots (a_1 - a_n)\} \{(a_2 - a_3)(a_2 - a_4) \dots (a_2 - a_n)\} \dots \{a_{n-1} - a_n\}$.

36. Without expanding, show that $\begin{vmatrix} na_1 + b_1 & na_2 + b_2 & na_3 + b_3 \\ nb_1 + c_1 & nb_2 + c_2 & nb_3 + c_3 \\ nc_1 + a_1 & nc_2 + a_2 & nc_3 + a_3 \end{vmatrix} = (n+1)(n^2 - n + 1) \begin{vmatrix} a_1 & a_2 & a_3 \\ b_1 & b_2 & b_3 \\ c_1 & c_2 & c_3 \end{vmatrix}$.

37. Without expanding, show that the equation $\begin{vmatrix} 0 & x-a & x-b \\ x+a & 0 & x-c \\ x+b & x+c & 0 \end{vmatrix} = 0$ has 0 as a root.

38. Prove $\begin{vmatrix} a+b & a & a & \dots & a \\ a & a+b & a & \dots & a \\ \dots & \dots & \dots & \dots & \dots \\ a & a & a & \dots & a+b \end{vmatrix} = b^{n-1}(na+b)$.

Chapter 4

Evaluation of Determinants

PROCEDURES FOR EVALUATING determinants of orders two and three are found in Chapter 3. In Problem 11 of that chapter, two uses of Theorem IX were illustrated: (a) to obtain an element 1 or -1 if the given determinant contains no such element, (b) to replace an element of a given determinant with 0.

For determinants of higher orders, the general procedure is to replace, by repeated use of Theorem IX, Chapter 3, the given determinant $|A|$ by another $|B| = |b_{ij}|$ having the property that all elements, except one, in some row (column) are zero. If b_{pq} is this non-zero element and β_{pq} is its cofactor,

$$|A| = |B| = b_{pq} \cdot \beta_{pq} = (-1)^{p+q} b_{pq} \cdot \text{minor of } b_{pq}$$

Then the minor of b_{pq} is treated in similar fashion and the process is continued until a determinant of order two or three is obtained.

Example 1.

$$\begin{vmatrix} 2 & 3 & -2 & 4 \\ 3 & -2 & 1 & 2 \\ 3 & 2 & 3 & 4 \\ -2 & 4 & 0 & 5 \end{vmatrix} = \begin{vmatrix} 2+2(3) & 3+2(-2) & -2+2(1) & 4+2(2) \\ 3 & -2 & 1 & 2 \\ 3-3(3) & 2-3(-2) & 3-3(1) & 4-3(2) \\ -2 & 4 & 0 & 5 \end{vmatrix} = \begin{vmatrix} 8 & -1 & 0 & 8 \\ 3 & -2 & 1 & 2 \\ -6 & 8 & 0 & -2 \\ -2 & 4 & 0 & 5 \end{vmatrix} \\ = (-1)^{2+3} \begin{vmatrix} 8 & -1 & 8 \\ -6 & 8 & -2 \\ -2 & 4 & 5 \end{vmatrix} = - \begin{vmatrix} 8+8(-1) & -1 & 8+8(-1) \\ -6+8(8) & 8 & -2+8(8) \\ -2+8(4) & 4 & 5+8(4) \end{vmatrix} = - \begin{vmatrix} 0 & -1 & 0 \\ 58 & 8 & 62 \\ 30 & 4 & 37 \end{vmatrix} \\ = -(-1)^{1+2}(-1) \begin{vmatrix} 58 & 62 \\ 30 & 37 \end{vmatrix} = -286$$

See Problems 1-3

For determinants having elements of the type in Example 2 below, the following variation may be used: divide the first row by one of its non-zero elements and proceed to obtain zero elements in a row or column.

Example 2.

$$\begin{vmatrix} 0.921 & 0.185 & 0.476 & 0.614 \\ 0.782 & 0.157 & 0.527 & 0.138 \\ 0.872 & 0.484 & 0.637 & 0.799 \\ 0.312 & 0.555 & 0.841 & 0.448 \end{vmatrix} = 0.921 \begin{vmatrix} 1 & 0.201 & 0.517 & 0.667 \\ 0.782 & 0.157 & 0.527 & 0.138 \\ 0.872 & 0.484 & 0.637 & 0.799 \\ 0.312 & 0.555 & 0.841 & 0.448 \end{vmatrix} = 0.921 \begin{vmatrix} 1 & 0.201 & 0.517 & 0.667 \\ 0 & 0 & 0.123 & -0.384 \\ 0 & 0.309 & 0.196 & 0.217 \\ 0 & 0.492 & 0.680 & 0.240 \end{vmatrix} \\ = 0.921 \begin{vmatrix} 0 & 0.123 & -0.384 \\ 0.309 & 0.196 & 0.217 \\ 0.492 & 0.680 & 0.240 \end{vmatrix} = 0.921(-0.384) \begin{vmatrix} 0 & -0.320 & 1 \\ 0.309 & 0.196 & 0.217 \\ 0.492 & 0.680 & 0.240 \end{vmatrix} \\ = 0.921(-0.384) \begin{vmatrix} 0 & 0 & 1 \\ 0.309 & 0.265 & 0.217 \\ 0.492 & 0.757 & 0.240 \end{vmatrix} = 0.921(-0.384) \begin{vmatrix} 0.309 & 0.265 \\ 0.492 & 0.757 \end{vmatrix} \\ = 0.921(-0.384)(0.104) = -0.037$$

THE LAPLACE EXPANSION. The expansion of a determinant $|A|$ of order n along a row (column) is a special case of the Laplace expansion. Instead of selecting one row of $|A|$, let m rows numbered i_1, i_2, \dots, i_m , when arranged in order of magnitude, be selected. From these m rows

$$\rho = \frac{n(n-1)\dots(n-m+1)}{1 \cdot 2 \dots m} \text{ minors } \left| A_{i_1, i_2, \dots, i_m}^{j_1, j_2, \dots, j_m} \right|$$

can be formed by making all possible selections of m columns from the n columns. Using these minors and their algebraic complements, we have the Laplace expansion

$$(4.1) \quad |A| = \sum_{\rho} (-1)^s \left| A_{i_1, i_2, \dots, i_m}^{j_1, j_2, \dots, j_m} \right| \cdot \left| A_{i_{m+1}, i_{m+2}, \dots, i_n}^{j_{m+1}, j_{m+2}, \dots, j_n} \right|$$

where $s = i_1 + i_2 + \dots + i_m + j_1 + j_2 + \dots + j_m$ and the summation extends over the ρ selections of the column indices taken m at a time.

Example 3.

$$\text{Evaluate } |A| = \begin{vmatrix} 2 & 3 & -2 & 4 \\ 3 & -2 & 1 & 2 \\ 3 & 2 & 3 & 4 \\ -2 & 4 & 0 & 5 \end{vmatrix} \text{ using minors of the first two rows.}$$

From (4.1),

$$\begin{aligned} |A| &= (-1)^{1+2+1+2} |A_{1,2}^{1,2}| \cdot |A_{3,4}^{3,4}| + (-1)^{1+2+1+3} |A_{1,2}^{1,3}| \cdot |A_{3,4}^{2,4}| \\ &\quad + (-1)^{1+2+1+4} |A_{1,2}^{1,4}| \cdot |A_{3,4}^{2,3}| + (-1)^{1+2+2+3} |A_{1,2}^{2,3}| \cdot |A_{3,4}^{1,4}| \\ &\quad + (-1)^{1+2+2+4} |A_{1,2}^{2,4}| \cdot |A_{3,4}^{1,3}| + (-1)^{1+2+3+4} |A_{1,2}^{3,4}| \cdot |A_{3,4}^{1,2}| \\ &= \begin{vmatrix} 2 & 3 \\ 3 & -2 \end{vmatrix} \cdot \begin{vmatrix} 3 & 4 \\ 0 & 5 \end{vmatrix} - \begin{vmatrix} 2 & -2 \\ 3 & 1 \end{vmatrix} \cdot \begin{vmatrix} 2 & 4 \\ 4 & 5 \end{vmatrix} + \begin{vmatrix} 2 & 4 \\ 3 & 2 \end{vmatrix} \cdot \begin{vmatrix} 2 & 3 \\ 4 & 0 \end{vmatrix} \\ &\quad + \begin{vmatrix} 3 & -2 \\ -2 & 1 \end{vmatrix} \cdot \begin{vmatrix} 3 & 4 \\ -2 & 5 \end{vmatrix} - \begin{vmatrix} 3 & 4 \\ -2 & 2 \end{vmatrix} \cdot \begin{vmatrix} 3 & 3 \\ -2 & 0 \end{vmatrix} + \begin{vmatrix} -2 & 4 \\ 1 & 2 \end{vmatrix} \cdot \begin{vmatrix} 3 & 2 \\ -2 & 4 \end{vmatrix} \\ &= (-13)(15) - (8)(-6) + (-8)(-12) + (-1)(23) - (14)(6) + (-8)(16) \\ &= -286 \end{aligned}$$

See Problems 4-6

DETERMINANT OF A PRODUCT. If A and B are n -square matrices, then

$$(4.2) \quad |AB| = |A| \cdot |B|$$

See Problem 7

EXPANSION ALONG THE FIRST ROW AND COLUMN. If $A = [a_{ij}]$ is n -square, then

$$(4.3) \quad |A| = a_{11}\alpha_{11} - \sum_{i=2}^n \sum_{j=2}^n a_{i1}a_{1j}\alpha_{ij}^{i1}$$

where α_{11} is the cofactor of a_{11} and α_{ij}^{i1} is the algebraic complement of the minor $\begin{vmatrix} a_{11} & a_{1j} \\ a_{i1} & a_{ij} \end{vmatrix}$ of A .

DERIVATIVE OF A DETERMINANT. Let the n -square matrix $A = [a_{ij}]$ have as elements differentiable functions of a variable x . Then

I. The derivative, $\frac{d}{dx}|A|$, of $|A|$ with respect to x is the sum of n determinants obtained by replacing in all possible ways the elements of one row (column) of $|A|$ by their derivatives with respect to x .

Example 4.

$$\begin{aligned} \frac{d}{dx} \begin{vmatrix} x^2 & x+1 & 3 \\ 1 & 2x-1 & x^3 \\ 0 & x & -2 \end{vmatrix} &= \begin{vmatrix} 2x & 1 & 0 \\ 1 & 2x-1 & x^3 \\ 0 & x & -2 \end{vmatrix} + \begin{vmatrix} x^2 & x+1 & 3 \\ 0 & 2 & 3x^2 \\ 0 & x & -2 \end{vmatrix} + \begin{vmatrix} x^2 & x+1 & 3 \\ 1 & 2x-1 & x^3 \\ 0 & 1 & 0 \end{vmatrix} \\ &= 5 + 4x - 12x^2 - 6x^5 \end{aligned}$$

See Problem 8

SOLVED PROBLEMS

$$1. \begin{vmatrix} 2 & 3 & -2 & 4 \\ 7 & 4 & -3 & 10 \\ 3 & 2 & 3 & 4 \\ -2 & 4 & 0 & 5 \end{vmatrix} = \begin{vmatrix} 2 & 3 & -2 & 4 \\ 7-2(2) & 4-2(3) & -3-2(-2) & 10-2(4) \\ 3 & 2 & 3 & 4 \\ -2 & 4 & 0 & 5 \end{vmatrix} = \begin{vmatrix} 2 & 3 & -2 & 4 \\ 3 & -2 & 1 & 2 \\ 3 & 2 & 3 & 4 \\ -2 & 4 & 0 & 5 \end{vmatrix} = -286 \text{ (See Example 1)}$$

There are, of course, many other ways of obtaining an element +1 or -1; for example, subtract the first column from the second, the fourth column from the second, the first row from the second, etc.

$$\begin{aligned} 2. \begin{vmatrix} 1 & 0 & -1 & 2 \\ 2 & 3 & 2 & -2 \\ 2 & 4 & 2 & 1 \\ 3 & 1 & 5 & -3 \end{vmatrix} &= \begin{vmatrix} 1 & 0 & -1+1 & 2-2(1) \\ 2 & 3 & 2+2 & -2-2(2) \\ 2 & 4 & 2+2 & 1-2(2) \\ 3 & 1 & 5+3 & -3-2(3) \end{vmatrix} = \begin{vmatrix} 1 & 0 & 0 & 0 \\ 2 & 3 & 4 & -6 \\ 2 & 4 & 4 & -3 \\ -3 & 1 & 8 & -9 \end{vmatrix} \\ &= \begin{vmatrix} 3 & 4 & -6 \\ 4 & 4 & -3 \\ 1 & 8 & -9 \end{vmatrix} = \begin{vmatrix} 3-2(4) & 4-2(4) & -6-2(-3) \\ 4 & 4 & -3 \\ 1-3(4) & 8-3(4) & -9-3(-3) \end{vmatrix} = \begin{vmatrix} -5 & -4 & 0 \\ 4 & 4 & -3 \\ -11 & -4 & 0 \end{vmatrix} \\ &= 3 \begin{vmatrix} -5 & -4 \\ -11 & -4 \end{vmatrix} = -72 \end{aligned}$$

$$3. \text{ Evaluate } |A| = \begin{vmatrix} 0 & 1+i & 1+2i \\ 1-i & 0 & 2-3i \\ 1-2i & 2+3i & 0 \end{vmatrix}$$

Multiply the second row by $1+i$ and the third row by $1+2i$; then

$$\begin{aligned} (1+i)(1+2i)|A| &= (-1+3i)|A| = \begin{vmatrix} 0 & 1+i & 1+2i \\ 2 & 0 & 5-i \\ 5 & -4+7i & 0 \end{vmatrix} = \begin{vmatrix} 0 & 1+i & 1+2i \\ 2 & 0 & 5-i \\ 1 & -4+7i & -10+2i \end{vmatrix} = \begin{vmatrix} 0 & 1+i & 1+2i \\ 0 & 8-14i & 25-5i \\ 1 & -4+7i & -10+2i \end{vmatrix} \\ &= \begin{vmatrix} 1+i & 1+2i \\ 8-14i & 25-5i \end{vmatrix} = -6 + 18i \end{aligned}$$

and $|A| = 6$.

4. Derive the Laplace expansion of $|A| = |a_{ij}|$ of order n , using minors of order $m < n$.

Consider the m -square minor $\begin{vmatrix} j_1, j_2, \dots, j_m \\ A_{i_1, i_2, \dots, i_m} \end{vmatrix}$ of $|A|$ in which the row and column indices are arranged in order of magnitude. Now by $i_1 - 1$ interchanges of adjacent rows of $|A|$, the row numbered i_1 can be brought into the first row, by $i_2 - 2$ interchanges of adjacent rows the row numbered i_2 can be brought into the second row, ..., by $i_m - m$ interchanges of adjacent rows the row numbered i_m can be brought into the m th row. Thus, after $(i_1 - 1) + (i_2 - 2) + \dots + (i_m - m) = i_1 + i_2 + \dots + i_m - \frac{1}{2}m(m+1)$ interchanges of adjacent rows the rows numbered i_1, i_2, \dots, i_m occupy the position of the first m rows. Similarly, after $j_1 + j_2 + \dots + j_m - \frac{1}{2}m(m+1)$ interchanges of adjacent columns, the columns numbered j_1, j_2, \dots, j_m occupy the position of the first m columns. As a result of the interchanges of adjacent rows and adjacent columns, the minor selected above occupies the upper left corner and its complement occupies the lower right corner of the determinant; moreover, $|A|$ has changed sign $\sigma = i_1 + i_2 + \dots + i_m + j_1 + j_2 + \dots + j_m - m(m+1)$ times which is equivalent to $s = i_1 + i_2 + \dots + i_m + j_1 + j_2 + \dots + j_m$ changes. Thus

$$\begin{aligned} & \begin{vmatrix} j_1, j_2, \dots, j_m \\ A_{i_1, i_2, \dots, i_m} \end{vmatrix} \cdot \begin{vmatrix} j_{m+1}, j_{m+2}, \dots, j_n \\ A_{i_{m+1}, i_{m+2}, \dots, i_n} \end{vmatrix} \text{ yields } m!(n-m)! \text{ terms of } (-1)^s |A| \text{ or} \\ (a) \quad & (-1)^s \begin{vmatrix} j_1, j_2, \dots, j_m \\ A_{i_1, i_2, \dots, i_m} \end{vmatrix} \cdot \begin{vmatrix} j_{m+1}, j_{m+2}, \dots, j_n \\ A_{i_{m+1}, i_{m+2}, \dots, i_n} \end{vmatrix} \text{ yields } m!(n-m)! \text{ terms of } |A|. \end{aligned}$$

Let i_1, i_2, \dots, i_m be held fixed. From these rows $\rho = \frac{n(n-1)\dots(n-m+1)}{1 \cdot 2 \dots m} = \frac{n!}{m!(n-m)!}$ different m -square minors may be selected. Each of these minors when multiplied by its algebraic complement yields $m!(n-m)!$ terms of $|A|$. Since, by their formation, there are no duplicate terms of $|A|$ among these products,

$$|A| = \sum_{\rho} (-1)^s \begin{vmatrix} j_1, j_2, \dots, j_m \\ A_{i_1, i_2, \dots, i_m} \end{vmatrix} \cdot \begin{vmatrix} j_{m+1}, j_{m+2}, \dots, j_n \\ A_{i_{m+1}, i_{m+2}, \dots, i_n} \end{vmatrix}$$

where $s = i_1 + i_2 + \dots + i_m + j_1 + j_2 + \dots + j_m$ and the summation extends over the ρ different selections j_1, j_2, \dots, j_m of the column indices.

5. Evaluate $|A| = \begin{vmatrix} 1 & 2 & 3 & 4 \\ 2 & 1 & 2 & 1 \\ 0 & 0 & 1 & 1 \\ 3 & 4 & 1 & 2 \end{vmatrix}$, using minors of the first two columns.

$$\begin{aligned} |A| &= (-1)^{1+2+1+2} \begin{vmatrix} 1 & 2 \\ 2 & 1 \end{vmatrix} \cdot \begin{vmatrix} 1 & 1 \\ 1 & 2 \end{vmatrix} + (-1)^{1+4+1+2} \begin{vmatrix} 1 & 2 \\ 3 & 4 \end{vmatrix} \cdot \begin{vmatrix} 2 & 1 \\ 1 & 1 \end{vmatrix} + (-1)^{2+4+1+2} \begin{vmatrix} 2 & 1 \\ 3 & 4 \end{vmatrix} \cdot \begin{vmatrix} 3 & 4 \\ 1 & 1 \end{vmatrix} \\ &= (-3)(1) + (-2)(1) - (5)(-1) \\ &= 0 \end{aligned}$$

6. If A, B , and C are n -square matrices, prove

$$|P| = \begin{vmatrix} A & O \\ C & B \end{vmatrix} = |A| \cdot |B|$$

From the first n rows of $|P|$ only one non-zero n -square minor, $|A|$, can be formed. Its algebraic complement is $|B|$. Hence, by the Laplace expansion, $|P| = |A| \cdot |B|$.

7. Prove $|AB| = |A| \cdot |B|$.

Suppose $A = [a_{ij}]$ and $B = [b_{ij}]$ are n -square. Let $C = [c_{ij}] = AB$ so that $c_{ij} = \sum_{k=1}^n a_{ik} b_{kj}$. From Problem 6,

$$|P| = \begin{vmatrix} a_{11} & a_{12} & \cdots & a_{1n} & 0 & 0 & \cdots & 0 \\ a_{21} & a_{22} & \cdots & a_{2n} & 0 & 0 & \cdots & 0 \\ \cdots & \cdots & \cdots & \cdots & \cdots & \cdots & \cdots & \cdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} & 0 & 0 & \cdots & 0 \\ -1 & 0 & \cdots & 0 & b_{11} & b_{12} & \cdots & b_{1n} \\ 0 & -1 & \cdots & 0 & b_{21} & b_{22} & \cdots & b_{2n} \\ \cdots & \cdots & \cdots & \cdots & \cdots & \cdots & \cdots & \cdots \\ 0 & 0 & \cdots & -1 & b_{n1} & b_{n2} & \cdots & b_{nn} \end{vmatrix} = |A| \cdot |B|$$

To the $(n+1)$ st column of $|P|$ add b_{11} times the first column, b_{21} times the second column, ..., b_{n1} times the n th column; we have

$$|P| = \begin{vmatrix} a_{11} & a_{12} & \cdots & a_{1n} & c_{11} & 0 & \cdots & 0 \\ a_{21} & a_{22} & \cdots & a_{2n} & c_{21} & 0 & \cdots & 0 \\ \cdots & \cdots & \cdots & \cdots & \cdots & \cdots & \cdots & \cdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} & c_{n1} & 0 & \cdots & 0 \\ -1 & 0 & \cdots & 0 & 0 & b_{12} & \cdots & b_{1n} \\ 0 & -1 & \cdots & 0 & 0 & b_{22} & \cdots & b_{2n} \\ \cdots & \cdots & \cdots & \cdots & \cdots & \cdots & \cdots & \cdots \\ 0 & 0 & \cdots & -1 & 0 & b_{n2} & \cdots & b_{nn} \end{vmatrix}$$

Next, to the $(n+2)$ nd column of $|P|$ add b_{12} times the first column, b_{22} times the second column, ..., b_{n2} times the n th column. We have

$$|P| = \begin{vmatrix} a_{11} & a_{12} & \cdots & a_{1n} & c_{11} & c_{12} & 0 & \cdots & 0 \\ a_{21} & a_{22} & \cdots & a_{2n} & c_{21} & c_{22} & 0 & \cdots & 0 \\ \cdots & \cdots & \cdots & \cdots & \cdots & \cdots & \cdots & \cdots & \cdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} & c_{n1} & c_{n2} & 0 & \cdots & 0 \\ -1 & 0 & \cdots & 0 & 0 & 0 & b_{13} & \cdots & b_{1n} \\ 0 & -1 & \cdots & 0 & 0 & 0 & b_{23} & \cdots & b_{2n} \\ \cdots & \cdots & \cdots & \cdots & \cdots & \cdots & \cdots & \cdots & \cdots \\ 0 & 0 & \cdots & -1 & 0 & 0 & b_{n3} & \cdots & b_{nn} \end{vmatrix}$$

Continuing this process, we obtain finally $|P| = \begin{vmatrix} A & C \\ -I_n & 0 \end{vmatrix}$. From the last n rows of $|P|$ only one non-zero n -square minor, $|-I_n| = (-1)^n$ can be formed. Its algebraic complement is $(-1)^{1+2+\cdots+n+(n+1)+\cdots+2n}|C| = (-1)^{n(2n+1)}|C|$. Hence, $|P| = (-1)^n(-1)^{n(2n+1)}|C| = |C|$ and $|C| = |AB| = |A| \cdot |B|$.

8. Let $A = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix}$ where $a_{ij} = a_{ij}(x)$, $(i, j = 1, 2, 3)$, are differentiable functions of x . Then

$$|A| = a_{11}a_{22}a_{33} + a_{12}a_{23}a_{31} + a_{13}a_{32}a_{21} - a_{11}a_{23}a_{32} - a_{12}a_{21}a_{33} - a_{13}a_{22}a_{31}$$

and, denoting $\frac{d}{dx} a_{ij}$ by a'_{ij} ,

$$\begin{aligned} \frac{d}{dx} |A| &= a'_{11}a_{22}a_{33} + a'_{22}a_{11}a_{33} + a'_{33}a_{11}a_{22} + a'_{12}a_{23}a_{31} + a'_{23}a_{12}a_{31} + a'_{31}a_{12}a_{23} \\ &\quad + a'_{13}a_{32}a_{21} + a'_{32}a_{13}a_{21} + a'_{21}a_{13}a_{32} - a'_{11}a_{23}a_{32} - a'_{23}a_{11}a_{32} - a'_{32}a_{11}a_{23} \\ &\quad - a'_{12}a_{21}a_{33} - a'_{21}a_{12}a_{33} - a'_{33}a_{12}a_{21} - a'_{13}a_{22}a_{31} - a'_{22}a_{13}a_{31} - a'_{31}a_{13}a_{22} \\ &= a'_{11}\alpha_{11} + a'_{12}\alpha_{12} + a'_{13}\alpha_{13} + a'_{21}\alpha_{21} + a'_{22}\alpha_{22} + a'_{23}\alpha_{23} + a'_{31}\alpha_{31} + a'_{32}\alpha_{32} + a'_{33}\alpha_{33} \\ &= \begin{vmatrix} a'_{11} & a'_{12} & a'_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{vmatrix} + \begin{vmatrix} a_{11} & a_{12} & a_{13} \\ a'_{21} & a'_{22} & a'_{23} \\ a_{31} & a_{32} & a_{33} \end{vmatrix} + \begin{vmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a'_{31} & a'_{32} & a'_{33} \end{vmatrix} \end{aligned}$$

by Problem 10, Chapter 3.

SUPPLEMENTARY PROBLEMS

9. Evaluate:

$$(a) \begin{vmatrix} 3 & 5 & 7 & 2 \\ 2 & 4 & 1 & 1 \\ -2 & 0 & 0 & 0 \\ 1 & 1 & 3 & 4 \end{vmatrix} = 156$$

$$(c) \begin{vmatrix} 1 & -2 & 3 & -4 \\ 2 & -1 & 4 & -3 \\ 2 & 3 & -4 & -5 \\ 3 & -4 & 5 & 6 \end{vmatrix} = -304$$

$$(b) \begin{vmatrix} 1 & 1 & 1 & 6 \\ 2 & 4 & 1 & 6 \\ 4 & 1 & 2 & 9 \\ 2 & 4 & 2 & 7 \end{vmatrix} = 41$$

$$(d) \begin{vmatrix} 1 & -2 & 3 & -2 & -2 \\ 2 & -1 & 1 & 3 & 2 \\ 1 & 1 & 2 & 1 & 1 \\ 1 & -4 & -3 & -2 & -5 \\ 3 & -2 & 2 & 2 & -2 \end{vmatrix} = 118$$

10. If A is n -square, show that $|\overline{A}A|$ is real and non-negative.

11. Evaluate the determinant of Problem 9(a) using minors from the first two rows; also using minors from the first two columns.

12. (a) Let $A = \begin{bmatrix} a_1 & a_2 \\ -a_2 & a_1 \end{bmatrix}$ and $B = \begin{bmatrix} b_1 & b_2 \\ -b_2 & b_1 \end{bmatrix}$.

Use $|AB| = |A| \cdot |B|$ to show that $(a_1^2 + a_2^2)(b_1^2 + b_2^2) = (a_1b_1 - a_2b_2)^2 + (a_2b_1 + a_1b_2)^2$.

(b) Let $A = \begin{bmatrix} a_1 + ia_3 & a_2 + ia_4 \\ -a_2 + ia_4 & a_1 - ia_3 \end{bmatrix}$ and $B = \begin{bmatrix} b_1 + ib_3 & b_2 + ib_4 \\ -b_2 + ib_4 & b_1 - ib_3 \end{bmatrix}$.

Use $|AB| = |A| \cdot |B|$ to express $(a_1^2 + a_2^2 + a_3^2 + a_4^2)(b_1^2 + b_2^2 + b_3^2 + b_4^2)$ as a sum of four squares.

13. Evaluate $\begin{vmatrix} 0 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 2 & 1 \\ 0 & 0 & 0 & 3 & 2 & 1 \\ 0 & 0 & 4 & 3 & 2 & 1 \\ 0 & 5 & 4 & 3 & 2 & 1 \\ 6 & 5 & 4 & 3 & 2 & 1 \end{vmatrix}$ using minors from the first three rows. Ans. -720

14. Evaluate $\begin{vmatrix} 1 & 2 & 1 & 2 & 1 \\ 0 & 0 & 1 & 1 & 1 \\ 1 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 1 & 2 \\ 1 & 2 & 2 & 1 & 1 \end{vmatrix}$ using minors from the first two columns. Ans. 2

15. If A_1, A_2, \dots, A_S are square matrices, use the Laplace expansion to prove

$$|\text{diag}(A_1, A_2, \dots, A_S)| = |A_1| \cdot |A_2| \cdots |A_S|$$

16. Expand $\begin{vmatrix} a_1 & a_2 & a_3 & a_4 \\ b_1 & b_2 & b_3 & b_4 \\ a_1 & a_2 & a_3 & a_4 \\ b_1 & b_2 & b_3 & b_4 \end{vmatrix}$ using minors of the first two rows and show that

$$\begin{vmatrix} a_1 & a_2 \\ b_1 & b_2 \end{vmatrix} \cdot \begin{vmatrix} a_3 & a_4 \\ b_3 & b_4 \end{vmatrix} - \begin{vmatrix} a_1 & a_3 \\ b_1 & b_3 \end{vmatrix} \cdot \begin{vmatrix} a_2 & a_4 \\ b_2 & b_4 \end{vmatrix} + \begin{vmatrix} a_1 & a_4 \\ b_1 & b_4 \end{vmatrix} \cdot \begin{vmatrix} a_2 & a_3 \\ b_2 & b_3 \end{vmatrix} = 0$$

17. Use the Laplace expansion to show that the n -square determinant $\begin{vmatrix} 0 & A \\ B & C \end{vmatrix}$, where 0 is k -square, is zero when $k > \frac{1}{2}n$.

18. In $|A| = a_{11}\alpha_{11} + a_{12}\alpha_{12} + a_{13}\alpha_{13} + a_{14}\alpha_{14}$ expand each of the cofactors $\alpha_{12}, \alpha_{13}, \alpha_{14}$ along its first column to show

$$|A| = a_{11}\alpha_{11} - \sum_{i=2}^4 \sum_{j=2}^4 a_{i1} a_{1j} \alpha_{ij}^{i1}$$

where α_{ij}^{i1} is the algebraic complement of the minor $\begin{vmatrix} a_{11} & a_{1j} \\ a_{i1} & a_{ij} \end{vmatrix}$ of $|A|$.

19. If α_{ij} denotes the cofactor of a_{ij} in the n -square matrix $A = [a_{ij}]$, show that the bordered determinant

$$\begin{vmatrix} a_{11} & a_{12} & \cdots & a_{1n} & p_1 \\ a_{21} & a_{22} & \cdots & a_{2n} & p_2 \\ \cdots & \cdots & \cdots & \cdots & \cdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} & p_n \\ q_1 & q_2 & \cdots & q_n & 0 \end{vmatrix} = \begin{vmatrix} 0 & q_1 & q_2 & \cdots & q_n \\ p_1 & a_{11} & a_{12} & \cdots & a_{1n} \\ \cdots & \cdots & \cdots & \cdots & \cdots \\ p_n & a_{n1} & a_{n2} & \cdots & a_{nn} \end{vmatrix} = - \sum_{i=1}^n \sum_{j=1}^n p_i q_j \alpha_{ij}$$

Hint. Use (4.3).

20. For each of the determinants $|A|$, find the derivative.

(a) $\begin{vmatrix} x^2 & x^3 \\ 2x & 3x+1 \end{vmatrix}$ (b) $\begin{vmatrix} x & 1 & 2 \\ x^2 & 2x+1 & x^3 \\ 0 & 3x-2 & x^2+1 \end{vmatrix}$ (c) $\begin{vmatrix} x^2-1 & x-1 & 1 \\ x^4 & x^3 & 2x+5 \\ x+1 & x^2 & x \end{vmatrix}$

Ans. (a) $2x + 9x^2 - 8x^3$, (b) $1 - 6x + 21x^2 + 12x^3 - 15x^4$, (c) $6x^5 - 5x^4 - 28x^3 + 9x^2 + 20x - 2$

21. Prove: If A and B are real n -square matrices with A non-singular and if $H = A + iB$ is Hermitian, then

$$|H|^2 = |A|^2 \cdot |I + (A^{-1}B)^2|$$

Chapter 5

Equivalence

THE RANK OF A MATRIX. A non-zero matrix A is said to have **rank** r if at least one of its r -square minors is different from zero while every $(r+1)$ -square minor, if any, is zero. A zero matrix is said to have rank 0.

Example 1. The rank of $A = \begin{bmatrix} 1 & 2 & 3 \\ 2 & 3 & 4 \\ 3 & 5 & 7 \end{bmatrix}$ is $r=2$ since $\begin{vmatrix} 1 & 2 \\ 2 & 3 \end{vmatrix} = -1 \neq 0$ while $|A| = 0$.

See Problem 1.

An n -square matrix A is called **non-singular** if its rank $r=n$, that is, if $|A| \neq 0$. Otherwise, A is called **singular**. The matrix of Example 1 is singular.

From $|AB| = |A| \cdot |B|$ follows

I. The product of two or more non-singular n -square matrices is non-singular; the product of two or more n -square matrices is singular if at least one of the matrices is singular.

ELEMENTARY TRANSFORMATIONS. The following operations, called **elementary transformations**, on a matrix do not change either its order or its rank:

- (1) The interchange of the i th and j th rows, denoted by H_{ij} ;
The interchange of the i th and j th columns, denoted by K_{ij} .
- (2) The multiplication of every element of the i th row by a non-zero scalar k , denoted by $H_i(k)$;
The multiplication of every element of the i th column by a non-zero scalar k , denoted by $K_i(k)$.
- (3) The addition to the elements of the i th row of k , a scalar, times the corresponding elements of the j th row, denoted by $H_{ij}(k)$;
The addition to the elements of the i th column of k , a scalar, times the corresponding elements of the j th column, denoted by $K_{ij}(k)$.

The transformations H are called **elementary row transformations**; the transformations K are called **elementary column transformations**.

The elementary transformations, being precisely those performed on the rows (columns) of a determinant, need no elaboration. It is clear that an elementary transformation cannot alter the order of a matrix. In Problem 2, it is shown that an elementary transformation does not alter its rank.

THE INVERSE OF AN ELEMENTARY TRANSFORMATION. The inverse of an elementary transformation is an operation which undoes the effect of the elementary transformation; that is, after A has been subjected to one of the elementary transformations and then the resulting matrix has been subjected to the inverse of that elementary transformation, the final result is the matrix A .

Example 2. Let $A = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{bmatrix}$

The effect of the elementary row transformation $H_{21}(-2)$ is to produce $B = \begin{bmatrix} 1 & 2 & 3 \\ 2 & 1 & 0 \\ 7 & 8 & 9 \end{bmatrix}$.

The effect of the elementary row transformation $H_{21}(+2)$ on B is to produce A again. Thus, $H_{21}(-2)$ and $H_{21}(+2)$ are inverse elementary row transformations.

The inverse elementary transformations are:

$$\begin{aligned} (1') \quad H_{ij}^{-1} &= H_{ij} & K_{ij}^{-1} &= K_{ij} \\ (2') \quad H_i^{-1}(k) &= H_i(1/k) & K_i^{-1}(k) &= K_i(1/k) \\ (3') \quad H_{ij}^{-1}(k) &= H_{ij}(-k) & K_{ij}^{-1}(k) &= K_{ij}(-k) \end{aligned}$$

We have

II. The inverse of an elementary transformation is an elementary transformation of the same type.

EQUIVALENT MATRICES. Two matrices A and B are called **equivalent**, $A \sim B$, if one can be obtained from the other by a sequence of elementary transformations.

Equivalent matrices have the same order and the same rank.

Example 3. Applying in turn the elementary transformations $H_{21}(-2)$, $H_{31}(1)$, $H_{32}(-1)$.

$$A = \begin{bmatrix} 1 & 2 & -1 & 4 \\ 2 & 4 & 3 & 5 \\ -1 & -2 & 6 & -7 \end{bmatrix} \sim \begin{bmatrix} 1 & 2 & -1 & 4 \\ 0 & 0 & 5 & -3 \\ -1 & -2 & 6 & -7 \end{bmatrix} \sim \begin{bmatrix} 1 & 2 & -1 & 4 \\ 0 & 0 & 5 & -3 \\ 0 & 0 & 5 & -3 \end{bmatrix} \sim \begin{bmatrix} 1 & 2 & -1 & 4 \\ 0 & 0 & 5 & -3 \\ 0 & 0 & 0 & 0 \end{bmatrix} = B$$

Since all 3-square minors of B are zero while $\begin{vmatrix} -1 & 4 \\ 5 & -3 \end{vmatrix} \neq 0$, the rank of B is 2; hence,

the rank of A is 2. This procedure of obtaining from A an equivalent matrix B from which the rank is evident by inspection is to be compared with that of computing the various minors of A .

See Problem 3.

ROW EQUIVALENCE. If a matrix A is reduced to B by the use of elementary row transformations alone, B is said to be **row equivalent** to A and conversely. The matrices A and B of Example 3 are row equivalent.

Any non-zero matrix A of rank r is row equivalent to a **canonical matrix** C in which

- one or more elements of each of the first r rows are non-zero while all other rows have only zero elements.
- in the i th row, ($i = 1, 2, \dots, r$), the first non-zero element is 1; let the column in which this element stands be numbered j_i .
- $j_1 < j_2 < \dots < j_r$.
- the only non-zero element in the column numbered j_i , ($i = 1, 2, \dots, r$), is the element 1 of the i th row.

To reduce A to C , suppose j_1 is the number of the first non-zero column of A .

(i₁) If $a_{1j_1} \neq 0$, use $H_1(1/a_{1j_1})$ to reduce it to 1, when necessary.

(i₂) If $a_{ij_1} = 0$ but $a_{pj_1} \neq 0$, use H_{1p} and proceed as in (i₁).

(ii) Use row transformations of type (3) with appropriate multiples of the first row to obtain zeros elsewhere in the j_1 st column.

If non-zero elements of the resulting matrix B occur only in the first row, $B = C$. Otherwise, suppose j_2 is the number of the first column in which this does not occur. If $b_{2j_2} \neq 0$, use $H_2(1/b_{2j_2})$ as in (i₁); if $b_{2j_2} = 0$ but $b_{qj_2} \neq 0$, use H_{2q} and proceed as in (i₁). Then, as in (ii), clear the j_2 nd column of all other non-zero elements.

If non-zero elements of the resulting matrix occur only in the first two rows, we have C . Otherwise, the procedure is repeated until C is reached.

Example 4. The sequence of row transformations $H_{21}(-2), H_{31}(1); H_2(1/5); H_{12}(1), H_{32}(-5)$ applied to A of Example 3 yields

$$A = \begin{bmatrix} 1 & 2 & -1 & 4 \\ 2 & 4 & 3 & 5 \\ -1 & -2 & 6 & -7 \end{bmatrix} \sim \begin{bmatrix} 1 & 2 & -1 & 4 \\ 0 & 0 & 5 & -3 \\ 0 & 0 & 5 & -3 \end{bmatrix} \sim \begin{bmatrix} 1 & 2 & -1 & 4 \\ 0 & 0 & 1 & -3/5 \\ 0 & 0 & 5 & -3 \end{bmatrix} \sim \begin{bmatrix} 1 & 2 & 0 & 17/5 \\ 0 & 0 & 1 & -3/5 \\ 0 & 0 & 0 & 0 \end{bmatrix} = C$$

having the properties (a)-(d).

See Problem 4.

THE NORMAL FORM OF A MATRIX. By means of elementary transformations any matrix A of rank $r > 0$ can be reduced to one of the forms

$$(5.1) \quad I_r, \quad \begin{bmatrix} I_r & 0 \\ 0 & 0 \end{bmatrix}, \quad [I_r \ 0], \quad \begin{bmatrix} I_r \\ 0 \end{bmatrix}$$

called its **normal form**. A zero matrix is its own normal form.

Since both row and column transformations may be used here, the element 1 of the first row obtained in the section above can be moved into the first column. Then both the first row and first column can be cleared of other non-zero elements. Similarly, the element 1 of the second row can be brought into the second column, and so on.

For example, the sequence $H_{21}(-2), H_{31}(1), K_{21}(-2), K_{31}(1), K_{41}(-4), K_{23}, K_2(1/5), H_{32}(-1), K_{42}(3)$ applied to A of Example 3 yields $\begin{bmatrix} I_2 & 0 \\ 0 & 0 \end{bmatrix}$, the normal form.

See Problem 5.

ELEMENTARY MATRICES. The matrix which results when an elementary row (column) transformation is applied to the identity matrix I_n is called an **elementary row (column) matrix**. Here, an elementary matrix will be denoted by the symbol introduced to denote the elementary transformation which produces the matrix.

Example 5. Examples of elementary matrices obtained from $I_3 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$ are:

$$H_{12} = \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix} = K_{12}, \quad H_3(k) = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & k \end{bmatrix} = K_3(k), \quad H_{23}(k) = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & k \\ 0 & 0 & 1 \end{bmatrix} = K_{32}(k)$$

Every elementary matrix is non-singular. (Why?)

The effect of applying an elementary transformation to an $m \times n$ matrix A can be produced by multiplying A by an elementary matrix.

To effect a given elementary row transformation on A of order $m \times n$, apply the transformation to I_m to form the corresponding elementary matrix H and multiply A on the left by H .

To effect a given elementary column transformation on A , apply the transformation to I_n to form the corresponding elementary matrix K and multiply A on the right by K .

Example 6. When $A = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{bmatrix}$, $H_{13} \cdot A = \begin{bmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{bmatrix} \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{bmatrix} = \begin{bmatrix} 7 & 8 & 9 \\ 4 & 5 & 6 \\ 1 & 2 & 3 \end{bmatrix}$ interchanges the first and third

rows of A ; $AK_{13}(2) = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{bmatrix} \cdot \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 2 & 0 & 1 \end{bmatrix} = \begin{bmatrix} 7 & 2 & 3 \\ 16 & 5 & 6 \\ 25 & 8 & 9 \end{bmatrix}$ adds to the first column of A two times

the third column.

LET A AND B BE EQUIVALENT MATRICES. Let the elementary row and column matrices corresponding to the elementary row and column transformations which reduce A to B be designated as H_1, H_2, \dots, H_s ; K_1, K_2, \dots, K_t where H_1 is the first row transformation, H_2 is the second, ...; K_1 is the first column transformation, K_2 is the second, ... Then

$$(5.2) \quad H_s \dots H_2 \cdot H_1 \cdot A \cdot K_1 \cdot K_2 \dots K_t = PAQ = B$$

where

$$(5.3) \quad P = H_s \dots H_2 \cdot H_1 \quad \text{and} \quad Q = K_1 \cdot K_2 \dots K_t$$

We have

III. Two matrices A and B are equivalent if and only if there exist non-singular matrices P and Q defined in (5.3) such that $PAQ = B$.

Example 7. When $A = \begin{bmatrix} 1 & 2 & -1 & 2 \\ 2 & 5 & -2 & 3 \\ 1 & 2 & 1 & 2 \end{bmatrix}$, $H_{31}(-1) \cdot H_{21}(-2) \cdot A \cdot K_{21}(-2) \cdot K_{31}(1) \cdot K_{41}(-2) \cdot K_{42}(1) \cdot K_3(\frac{1}{2})$

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

$$= \begin{bmatrix} 1 & 0 & 0 \\ -2 & 1 & 0 \\ -1 & 0 & 1 \end{bmatrix} \cdot A \cdot \begin{bmatrix} 1 & -2 & \frac{1}{2} & -4 \\ 0 & 1 & 0 & 1 \\ 0 & 0 & \frac{1}{2} & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} = PAQ = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} = B$$

Since any matrix is equivalent to its normal form, we have

IV. If A is an n -square non-singular matrix, there exist non-singular matrices P and Q as defined in (5.3) such that $PAQ = I_n$.

See Problem 6.

INVERSE OF A PRODUCT OF ELEMENTARY MATRICES. Let

$$P = H_s \dots H_2 \cdot H_1 \quad \text{and} \quad Q = K_1 \cdot K_2 \dots K_t$$

as in (5.3). Since each H and K has an inverse and since the inverse of a product is the product in reverse order of the inverses of the factors

$$(5.4) \quad P^{-1} = H_1^{-1} \cdot H_2^{-1} \dots H_s^{-1} \quad \text{and} \quad Q^{-1} = K_t^{-1} \dots K_2^{-1} \cdot K_1^{-1}$$

Let A be an n -square non-singular matrix and let P and Q defined above be such that $PAQ = I_n$. Then

$$(5.5) \quad A = P^{-1}(PAQ)Q^{-1} = P^{-1} \cdot I_n \cdot Q^{-1} = P^{-1} \cdot Q^{-1}$$

We have proved.

V. Every non-singular matrix can be expressed as a product of elementary matrices.

See Problem 7.

From this follow

VI. If A is non-singular, the rank of AB (also of BA) is that of B .

VII. If P and Q are non-singular, the rank of PAQ is that of A .

CANONICAL SETS UNDER EQUIVALENCE. In Problem 8, we prove

VIII. Two $m \times n$ matrices A and B are equivalent if and only if they have the same rank.

A set of $m \times n$ matrices is called a **canonical set** under equivalence if every $m \times n$ matrix is equivalent to one and only one matrix of the set. Such a canonical set is given by (5.1) as r ranges over the values $1, 2, \dots, m$ or $1, 2, \dots, n$ whichever is the smaller.

See Problem 9.

RANK OF A PRODUCT. Let A be an $m \times p$ matrix of rank r . By Theorem III there exist non-singular matrices P and Q such that

$$PAQ = N = \begin{bmatrix} I_r & 0 \\ 0 & 0 \end{bmatrix}$$

Then $A = P^{-1}NQ^{-1}$. Let B be a $p \times n$ matrix and consider the rank of

$$(5.6) \quad AB = P^{-1}NQ^{-1}B$$

By Theorem VI, the rank of AB is that of $NQ^{-1}B$. Now the rows of $NQ^{-1}B$ consist of the first r rows of $Q^{-1}B$ and $m-r$ rows of zeros. Hence, the rank of AB cannot exceed r , the rank of A . Similarly, the rank of AB cannot exceed that of B . We have proved

IX. The rank of the product of two matrices cannot exceed the rank of either factor.

Suppose $AB = 0$; then from (5.6), $NQ^{-1}B = 0$. This requires that the first r rows of $Q^{-1}B$ be zeros while the remaining rows may be arbitrary. Thus, the rank of $Q^{-1}B$ and, hence, the rank of B cannot exceed $p-r$. We have proved

X. If the $m \times p$ matrix A is of rank r and if the $p \times n$ matrix B is such that $AB = 0$, the rank of B cannot exceed $p-r$.

SOLVED PROBLEMS

1. (a) The rank of $A = \begin{bmatrix} 1 & 2 & 3 \\ -4 & 0 & 5 \end{bmatrix}$ is 2 since $\begin{vmatrix} 1 & 2 \\ -4 & 0 \end{vmatrix} \neq 0$ and there are no minors of order three.

(b) The rank of $A = \begin{bmatrix} 1 & 2 & 3 \\ 1 & 2 & 5 \\ 2 & 4 & 8 \end{bmatrix}$ is 2 since $|A| = 0$ and $\begin{vmatrix} 2 & 3 \\ 2 & 5 \end{vmatrix} \neq 0$.

(c) The rank of $A = \begin{bmatrix} 0 & 2 & 3 \\ 0 & 4 & 6 \\ 0 & 6 & 9 \end{bmatrix}$ is 1 since $|A| = 0$, each of the nine 2-square minors is 0, but not every element is 0.

2. Show that the elementary transformations do not alter the rank of a matrix.

We shall consider only row transformations here and leave consideration of the column transformations as an exercise. Let the rank of the $m \times n$ matrix A be r so that every $(r+1)$ -square minor of A , if any, is zero. Let B be the matrix obtained from A by a row transformation. Denote by $|R|$ any $(r+1)$ -square minor of A and by $|S|$ the $(r+1)$ -square minor of B having the same position as $|R|$.

Let the row transformation be H_{ij} . Its effect on $|R|$ is either (i) to leave it unchanged, (ii) to interchange two of its rows, or (iii) to interchange one of its rows with a row not of $|R|$. In the case (i), $|S| = |R| = 0$; in the case (ii), $|S| = -|R| = 0$; in the case (iii), $|S|$ is, except possibly for sign, another $(r+1)$ -square minor of $|A|$ and, hence, is 0.

Let the row transformation be $H_i(k)$. Its effect on $|R|$ is either (i) to leave it unchanged or (ii) to multiply one of its rows by k . Then, respectively, $|S| = |R| = 0$ or $|S| = k|R| = 0$.

Let the row transformation be $H_{ij}(k)$. Its effect on $|R|$ is either (i) to leave it unchanged, (ii) to increase one of its rows by k times another of its rows, or (iii) to increase one of its rows by k times a row not of $|R|$. In the cases (i) and (ii), $|S| = |R| = 0$; in the case (iii), $|S| = |R| \pm k$ (another $(r+1)$ -square minor of A) = $0 \pm k \cdot 0 = 0$.

Thus, an elementary row transformation cannot raise the rank of a matrix. On the other hand, it cannot lower the rank for, if it did, the inverse transformation would have to raise it. Hence, an elementary row transformation does not alter the rank of a matrix.

3. For each of the matrices A obtain an equivalent matrix B and from it, by inspection, determine the rank of A .

$$(a) A = \begin{bmatrix} 1 & 2 & 3 \\ 2 & 1 & 3 \\ 3 & 2 & 1 \end{bmatrix} \sim \begin{bmatrix} 1 & 2 & 3 \\ 0 & -3 & -3 \\ 0 & -4 & -8 \end{bmatrix} \sim \begin{bmatrix} 1 & 2 & 3 \\ 0 & 1 & 1 \\ 0 & 1 & 2 \end{bmatrix} \sim \begin{bmatrix} 1 & 2 & 3 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{bmatrix} = B$$

The transformations used were $H_{21}(-2)$, $H_{31}(-3)$; $H_2(-1/3)$, $H_3(-1/4)$; $H_{32}(-1)$. The rank is 3.

$$(b) A = \begin{bmatrix} 1 & 2 & 3 & 0 \\ 2 & 4 & 3 & 2 \\ 3 & 2 & 1 & 3 \\ 6 & 8 & 7 & 5 \end{bmatrix} \sim \begin{bmatrix} 1 & 2 & 3 & 0 \\ 0 & 0 & -3 & 2 \\ 0 & -4 & -8 & 3 \\ 0 & -4 & -11 & 5 \end{bmatrix} \sim \begin{bmatrix} 1 & 2 & 3 & 0 \\ 0 & -4 & -8 & 3 \\ 0 & 0 & -3 & 2 \\ 0 & -4 & -11 & 5 \end{bmatrix} \sim \begin{bmatrix} 1 & 2 & 3 & 0 \\ 0 & -4 & -8 & 3 \\ 0 & 0 & -3 & 2 \\ 0 & 0 & -3 & 2 \end{bmatrix} \sim \begin{bmatrix} 1 & 2 & 3 & 0 \\ 0 & -4 & -8 & 3 \\ 0 & 0 & -3 & 2 \\ 0 & 0 & 0 & 0 \end{bmatrix} = B. \text{ The rank is 3.}$$

$$(c) A = \begin{bmatrix} 1 & 1+i & -i \\ 0 & i & 1+2i \\ 1 & 1+2i & 1+i \end{bmatrix} \sim \begin{bmatrix} 1 & 0 & 0 \\ 0 & i & 1+2i \\ 1 & i & 1+2i \end{bmatrix} \sim \begin{bmatrix} 1 & 0 & 0 \\ 0 & i & 1+2i \\ 0 & 0 & 0 \end{bmatrix} = B. \text{ The rank is 2.}$$

Note. The equivalent matrices B obtained here are not unique. In particular, since in (a) and (b) only row transformations were used, the reader may obtain others by using only column transformations. When the elements are rational numbers, there generally is no gain in mixing row and column transformations.

$$\text{Thus, } P_1 = \begin{bmatrix} 1 & 0 & 0 \\ -2 & 1 & 0 \\ -1 & -1 & 1 \end{bmatrix}, \quad Q_1 = \begin{bmatrix} 1 & 1/3 & -4/3 & -1/3 \\ 0 & -1/6 & -5/6 & 7/6 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \quad \text{and} \quad P_1 A Q_1 = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} = N.$$

7. Express $A = \begin{bmatrix} 1 & 3 & 3 \\ 1 & 4 & 3 \\ 1 & 3 & 4 \end{bmatrix}$ as a product of elementary matrices.

The elementary transformations $H_{21}(-1)$, $H_{31}(-1)$; $K_{21}(-3)$, $K_{31}(-3)$ reduce A to I_3 , that is, [see (5.2)]

$$I = H_2 \cdot H_1 \cdot A \cdot K_1 \cdot K_2 = H_{31}(-1) \cdot H_{21}(-1) \cdot A \cdot K_{21}(-3) \cdot K_{31}(-3)$$

$$\text{From (5.5), } A = H_1^{-1} \cdot H_2^{-1} \cdot K_2^{-1} \cdot K_1^{-1} = \begin{bmatrix} 1 & 0 & 0 \\ 1 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 1 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 3 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 3 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

8. Prove: Two $m \times n$ matrices A and B are equivalent if and only if they have the same rank.

If A and B have the same rank, both are equivalent to the same matrix (5.1) and are equivalent to each other. Conversely, if A and B are equivalent, there exist non-singular matrices P and Q such that $B = PAQ$. By Theorem VII, A and B have the same rank.

9. A canonical set for non-zero matrices of order 3 is

$$I_3, \quad \begin{bmatrix} I_2 & 0 \\ 0 & 0 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix}, \quad \begin{bmatrix} I_1 & 0 \\ 0 & 0 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

A canonical set for non-zero 3×4 -matrices is

$$[I_3 \ 0] = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix}, \quad \begin{bmatrix} I_2 & 0 \\ 0 & 0 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}, \quad \begin{bmatrix} I_1 & 0 \\ 0 & 0 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

10. If from a square matrix A of order n and rank r_A , a submatrix B consisting of s rows (columns) of A is selected, the rank r_B of B is equal to or greater than $r_A + s - n$.

The normal form of A has $n - r_A$ rows whose elements are zeros and the normal form of B has $s - r_B$ rows whose elements are zeros. Clearly

$$n - r_A \geq s - r_B$$

from which follows $r_B \geq r_A + s - n$ as required.

SUPPLEMENTARY PROBLEMS

11. Find the rank of (a) $\begin{bmatrix} 1 & 2 & 3 & 2 \\ 2 & 3 & 5 & 1 \\ 1 & 3 & 4 & 5 \end{bmatrix}$, (b) $\begin{bmatrix} 1 & 2 & 1 & 2 \\ 1 & 3 & 2 & 2 \\ 2 & 4 & 3 & 4 \\ 3 & 7 & 4 & 6 \end{bmatrix}$, (c) $\begin{bmatrix} 1 & 2 & -2 & 3 \\ 2 & 5 & -4 & 6 \\ -1 & -3 & 2 & -2 \\ 2 & 4 & -1 & 6 \end{bmatrix}$, (d) $\begin{bmatrix} 3 & 4 & 5 & 6 & 7 \\ 4 & 5 & 6 & 7 & 8 \\ 5 & 6 & 7 & 8 & 9 \\ 10 & 11 & 12 & 13 & 14 \\ 15 & 16 & 17 & 18 & 19 \end{bmatrix}$

Ans. (a) 2, (b) 3, (c) 4, (d) 2

12. Show by considering minors that A, A', \bar{A} , and \bar{A}' have the same rank.

13. Show that the canonical matrix C , row equivalent to a given matrix A , is uniquely determined by A .

14. Find the canonical matrix row equivalent to each of the following:

(a) $\begin{bmatrix} 1 & 2 & -3 \\ 2 & 5 & -4 \end{bmatrix} \sim \begin{bmatrix} 1 & 0 & -7 \\ 0 & 1 & 2 \end{bmatrix}$ (b) $\begin{bmatrix} 1 & 2 & 3 & 4 \\ 3 & 4 & 1 & 2 \\ 4 & 3 & 1 & 2 \end{bmatrix} \sim \begin{bmatrix} 1 & 0 & 0 & 1/9 \\ 0 & 1 & 0 & 1/9 \\ 0 & 0 & 1 & 11/9 \end{bmatrix}$ (c) $\begin{bmatrix} 1 & 1 & 1 & 2 \\ 2 & 1 & -3 & -6 \\ 3 & -3 & 1 & 2 \end{bmatrix} \sim \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 2 \end{bmatrix}$

(d) $\begin{bmatrix} 1 & 2 & 1 & 0 \\ 3 & 2 & 1 & 2 \\ 2 & -1 & 2 & 5 \\ 5 & 6 & 3 & 2 \\ 1 & 3 & -1 & -3 \end{bmatrix} \sim \begin{bmatrix} 1 & 0 & 0 & 1 \\ 0 & 1 & 0 & -1 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$ (e) $\begin{bmatrix} 1 & -1 & 1 & 1 & 1 \\ 1 & -1 & 2 & 3 & 1 \\ 2 & -2 & 1 & 0 & 2 \\ 1 & 1 & -1 & -3 & 3 \end{bmatrix} \sim \begin{bmatrix} 1 & 0 & 0 & -1 & 2 \\ 0 & 1 & 0 & 0 & 1 \\ 0 & 0 & 1 & 2 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}$

15. Write the normal form of each of the matrices of Problem 14.

Ans. (a) $[I_2 \ 0]$, (b), (c) $[I_3 \ 0]$ (d) $\begin{bmatrix} I_3 & 0 \\ 0 & 0 \end{bmatrix}$ (e) $\begin{bmatrix} I_3 & 0 \\ 0 & 0 \end{bmatrix}$

16. Let $A = \begin{bmatrix} 1 & 2 & 3 & 4 \\ 2 & 3 & 4 & 1 \\ 3 & 4 & 1 & 2 \end{bmatrix}$

- (a) From I_3 form $H_{12}, H_2(3), H_{13}(-4)$ and check that each HA effects the corresponding row transformation.
- (b) From I_4 form $K_{24}, K_3(-1), K_{42}(3)$ and show that each AK effects the corresponding column transformation.
- (c) Write the inverses $H_{12}^{-1}, H_2^{-1}(3), H_{13}^{-1}(-4)$ of the elementary matrices of (a). Check that for each $H, H \cdot H^{-1} = I$.
- (d) Write the inverses $K_{24}^{-1}, K_3^{-1}(-1), K_{42}^{-1}(3)$ of the elementary matrices of (b). Check that for each $K, KK^{-1} = I$.

(e) Compute $B = H_{12} \cdot H_2(3) \cdot H_{13}(-4) = \begin{bmatrix} 0 & 3 & 0 \\ 1 & 0 & -4 \\ 0 & 0 & 1 \end{bmatrix}$ and $C = H_{13}^{-1}(-4) \cdot H_2^{-1}(3) \cdot H_{12}^{-1} = \begin{bmatrix} 0 & 1 & 4 \\ 1/3 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}$.

(f) Show that $BC = CB = I$.

- 17. (a) Show that $K'_{ij} = H_{ij}, K'_i(k) = H_i(k)$, and $K'_{ij}(k) = H_{ij}(k)$.
- (b) Show that if R is a product of elementary column matrices, R' is the product in reverse order of the same elementary row matrices.

- 18. Prove: (a) AB and BA are non-singular if A and B are non-singular n -square matrices.
- (b) AB and BA are singular if at least one of the n -square matrices A and B is singular.

- 19. If P and Q are non-singular, show that A, PA, AQ , and PAQ have the same rank.
- Hint. Express P and Q as products of elementary matrices.

20. Reduce $B = \begin{bmatrix} 1 & 3 & 6 & -1 \\ 1 & 4 & 5 & 1 \\ 1 & 5 & 4 & 3 \end{bmatrix}$ to normal form N and compute the matrices P_2 and Q_2 such that $P_2 B Q_2 = N$.

21. (a) Show that the number of matrices in a canonical set of n -square matrices under equivalence is $n+1$.
 (b) Show that the number of matrices in a canonical set of $m \times n$ matrices under equivalence is the smaller of $m+1$ and $n+1$.

22. Given $A = \begin{bmatrix} 1 & 2 & 4 & 4 \\ 1 & 3 & 2 & 6 \\ 2 & 5 & 6 & 10 \end{bmatrix}$ of rank 2. Find a 4-square matrix $B \neq 0$ such that $AB = 0$.

Hint. Follow the proof of Theorem X and take

$$Q^{-1}B = \begin{bmatrix} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ a & b & c & d \\ e & f & g & h \end{bmatrix}$$

where a, b, \dots, h are arbitrary.

23. The matrix A of Problem 6 and the matrix B of Problem 20 are equivalent. Find P and Q such that $B = PAQ$.
24. If the $m \times n$ matrices A and B are of rank r_A and r_B respectively, show that the rank of $A+B$ cannot exceed $r_A + r_B$.
25. Let A be an arbitrary n -square matrix and B be an n -square elementary matrix. By considering each of the six different types of matrix B , show that $|AB| = |A| \cdot |B|$.
26. Let A and B be n -square matrices. (a) If at least one is singular show that $|AB| = |A| \cdot |B|$; (b) If both are non-singular, use (3.5) and Problem 25 to show that $|AB| = |A| \cdot |B|$.
27. Show that equivalence of matrices is an equivalence relation.
28. Prove: The row equivalent canonical form of a non-singular matrix A is I and conversely.
29. Prove: Not every matrix A can be reduced to normal form by row transformations alone.
Hint. Exhibit a matrix which cannot be so reduced.
30. Show how to effect on any matrix A the transformation H_{ij} by using a succession of row transformations of types (2) and (3).
31. Prove: If A is an $m \times n$ matrix, ($m \leq n$), of rank m then AA' is a non-singular symmetric matrix. State the theorem when the rank of A is $< m$.

Chapter 6

The Adjoint of a Square Matrix

THE ADJOINT. Let $A = [a_{ij}]$ be an n -square matrix and α_{ij} be the cofactor of a_{ij} ; then by definition,

$$(6.1) \quad \text{adjoint } A = \text{adj } A = \begin{bmatrix} \alpha_{11} & \alpha_{21} & \cdots & \alpha_{n1} \\ \alpha_{12} & \alpha_{22} & \cdots & \alpha_{n2} \\ \cdots & \cdots & \cdots & \cdots \\ \alpha_{1n} & \alpha_{2n} & \cdots & \alpha_{nn} \end{bmatrix}$$

Note carefully that the cofactors of the elements of the i th row (column) of A are the elements of the i th column (row) of $\text{adj } A$.

Example 1. For the matrix $A = \begin{bmatrix} 1 & 2 & 3 \\ 2 & 3 & 2 \\ 3 & 3 & 4 \end{bmatrix}$.

$$\alpha_{11} = 6, \quad \alpha_{12} = -2, \quad \alpha_{13} = -3, \quad \alpha_{21} = 1, \quad \alpha_{22} = -5, \quad \alpha_{23} = 3, \quad \alpha_{31} = -5, \quad \alpha_{32} = 4, \quad \alpha_{33} = -1$$

and
$$\text{adj } A = \begin{bmatrix} 6 & 1 & -5 \\ -2 & -5 & 4 \\ -3 & 3 & -1 \end{bmatrix}$$

See Problems 1-2.

Using Theorems X and XI of Chapter 3, we find

$$(6.2) \quad A(\text{adj } A) = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \cdots & \cdots & \cdots & \cdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{bmatrix} \begin{bmatrix} \alpha_{11} & \alpha_{21} & \cdots & \alpha_{n1} \\ \alpha_{12} & \alpha_{22} & \cdots & \alpha_{n2} \\ \cdots & \cdots & \cdots & \cdots \\ \alpha_{1n} & \alpha_{2n} & \cdots & \alpha_{nn} \end{bmatrix}$$

$$= \text{diag}(|A|, |A|, \dots, |A|) = |A| \cdot I_n = (\text{adj } A) A$$

Example 2. For the matrix A of Example 1, $|A| = -7$ and

$$A(\text{adj } A) = \begin{bmatrix} 1 & 2 & 3 \\ 2 & 3 & 2 \\ 3 & 3 & 4 \end{bmatrix} \begin{bmatrix} 6 & 1 & -5 \\ -2 & -5 & 4 \\ -3 & 3 & -1 \end{bmatrix} = \begin{bmatrix} -7 & 0 & 0 \\ 0 & -7 & 0 \\ 0 & 0 & -7 \end{bmatrix} = -7I$$

By taking determinants in (6.2), we have

$$(6.3) \quad |A| \cdot |\text{adj } A| = |A|^n = |\text{adj } A| \cdot |A|$$

There follow

I. If A is n -square and non-singular, then

$$(6.4) \quad |\text{adj } A| = |A|^{n-1}$$

II. If A is n -square and singular, then

$$A(\text{adj } A) = (\text{adj } A)A = 0$$

If A is of rank $< n-1$, then $\text{adj } A = 0$. If A is of rank $n-1$, then $\text{adj } A$ is of rank 1.

See Problem 3.

THE ADJOINT OF A PRODUCT. In Problem 4, we prove

III. If A and B are n -square matrices,

$$(6.5) \quad \text{adj } AB = \text{adj } B \cdot \text{adj } A$$

MINOR OF AN ADJOINT. In Problem 6, we prove

IV. Let $\begin{vmatrix} A_{i_1, i_2, \dots, i_m}^{j_1, j_2, \dots, j_m} \end{vmatrix}$ be an m -square minor of the n -square matrix $A = [a_{ij}]$,

let $\begin{vmatrix} A_{i_{m+1}, i_{m+2}, \dots, i_n}^{j_{m+1}, j_{m+2}, \dots, j_n} \end{vmatrix}$ be its complement in A , and

let $\begin{vmatrix} M_{i_1, i_2, \dots, i_m}^{j_1, j_2, \dots, j_m} \end{vmatrix}$ denote the m -square minor of $\text{adj } A$ whose elements occupy the same position in $\text{adj } A$ as those of $\begin{vmatrix} A_{i_1, i_2, \dots, i_m}^{j_1, j_2, \dots, j_m} \end{vmatrix}$ occupy in A .

Then

$$(6.6) \quad |A| \cdot \begin{vmatrix} M_{i_1, i_2, \dots, i_m}^{j_1, j_2, \dots, j_m} \end{vmatrix} = (-1)^s |A|^m \cdot \begin{vmatrix} A_{i_{m+1}, i_{m+2}, \dots, i_n}^{j_{m+1}, j_{m+2}, \dots, j_n} \end{vmatrix}$$

where $s = i_1 + i_2 + \dots + i_m + j_1 + j_2 + \dots + j_m$.

If in (6.6), A is non-singular, then

$$(6.7) \quad \begin{vmatrix} M_{i_1, i_2, \dots, i_m}^{j_1, j_2, \dots, j_m} \end{vmatrix} = (-1)^s |A|^{m-1} \cdot \begin{vmatrix} A_{i_{m+1}, i_{m+2}, \dots, i_n}^{j_{m+1}, j_{m+2}, \dots, j_n} \end{vmatrix}$$

When $m = 2$, (6.7) becomes

$$(6.8) \quad \begin{vmatrix} \alpha_{i_1, j_1} & \alpha_{i_2, j_1} \\ \alpha_{i_1, j_2} & \alpha_{i_2, j_2} \end{vmatrix} = (-1)^{i_1+i_2+j_1+j_2} |A| \cdot \begin{vmatrix} A_{i_3, i_4, \dots, i_n}^{j_3, j_4, \dots, j_n} \end{vmatrix} \\ = |A| \cdot \text{algebraic complement of } \begin{vmatrix} A_{i_1, i_2}^{j_1, j_2} \end{vmatrix}$$

When $m = n-1$, (6.7) becomes

$$(6.9) \quad \begin{vmatrix} M_{i_1, i_2, \dots, i_{n-1}}^{j_1, j_2, \dots, j_{n-1}} \end{vmatrix} = (-1)^{i_n+j_n} |A|^{n-2} a_{i_n, j_n}$$

When $m = n$, (6.7) becomes (6.4).

SOLVED PROBLEMS

1. The adjoint of $A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$ is $\begin{bmatrix} \alpha_{11} & \alpha_{21} \\ \alpha_{12} & \alpha_{22} \end{bmatrix} = \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$.

2. The adjoint of $A = \begin{bmatrix} 1 & 2 & 3 \\ 1 & 3 & 4 \\ 1 & 4 & 3 \end{bmatrix}$ is $\begin{bmatrix} \begin{vmatrix} 3 & 4 \\ 4 & 3 \end{vmatrix} & -\begin{vmatrix} 2 & 3 \\ 4 & 3 \end{vmatrix} & \begin{vmatrix} 2 & 3 \\ 3 & 4 \end{vmatrix} \\ -\begin{vmatrix} 1 & 4 \\ 1 & 3 \end{vmatrix} & \begin{vmatrix} 1 & 3 \\ 1 & 3 \end{vmatrix} & -\begin{vmatrix} 1 & 3 \\ 1 & 4 \end{vmatrix} \\ \begin{vmatrix} 1 & 3 \\ 1 & 4 \end{vmatrix} & -\begin{vmatrix} 1 & 2 \\ 1 & 4 \end{vmatrix} & \begin{vmatrix} 1 & 2 \\ 1 & 3 \end{vmatrix} \end{bmatrix} = \begin{bmatrix} -7 & 6 & -1 \\ 1 & 0 & -1 \\ 1 & -2 & 1 \end{bmatrix}$.

3. Prove: If A is of order n and rank $n - 1$, then $\text{adj } A$ is of rank 1.

First we note that, since A is of rank $n - 1$, there is at least one non-zero cofactor and the rank of $\text{adj } A$ is at least one. By Theorem X, Chapter 5, the rank of $\text{adj } A$ is at most $n - (n - 1) = 1$. Hence, the rank is exactly one.

4. Prove: $\text{adj } AB = \text{adj } B \cdot \text{adj } A$.

By (6.2) $AB \text{adj } AB = |AB| \cdot I = (\text{adj } AB) AB$

Since $AB \cdot \text{adj } B \cdot \text{adj } A = A(B \cdot \text{adj } B) \text{adj } A = A(|B| \cdot I) \text{adj } A = |B|(A \text{adj } A) = |B| \cdot |A| \cdot I = |AB| \cdot I$

and $(\text{adj } B \cdot \text{adj } A) AB = \text{adj } B \{(\text{adj } A) A\} B = \text{adj } B \cdot |A| \cdot I \cdot B = |A| \{(\text{adj } B) B\} = |AB| \cdot I$

we conclude that $\text{adj } AB = \text{adj } B \cdot \text{adj } A$

5. Show that $\text{adj}(\text{adj } A) = |A|^{n-2} \cdot A$, if $|A| \neq 0$.

By (6.2) and (6.4),

$$\begin{aligned} \text{adj } A \cdot \text{adj}(\text{adj } A) &= \text{diag}(|\text{adj } A|, |\text{adj } A|, \dots, |\text{adj } A|) \\ &= \text{diag}(|A|^{n-1}, |A|^{n-1}, \dots, |A|^{n-1}) \end{aligned}$$

Then

$$A \cdot \text{adj } A \cdot \text{adj}(\text{adj } A) = |A|^{n-1} \cdot A$$

$$|A| \cdot \text{adj}(\text{adj } A) = |A|^{n-1} \cdot A$$

and

$$\text{adj}(\text{adj } A) = |A|^{n-2} \cdot A$$

6. Prove: Let $\begin{vmatrix} j_1 & j_2 & \dots & j_m \\ A_{i_1} & i_2 & \dots & i_m \end{vmatrix}$ be an m -square minor of the n -square matrix $A = [a_{ij}]$,

let $\begin{vmatrix} j_{m+1} & j_{m+2} & \dots & j_n \\ A_{i_{m+1}} & i_{m+2} & \dots & i_n \end{vmatrix}$ be its complement in A , and

let $\begin{vmatrix} j_1 & j_2 & \dots & j_m \\ M_{i_1} & i_2 & \dots & i_m \end{vmatrix}$ denote the m -square minor of $\text{adj } A$ whose elements occupy the same

position in $\text{adj } A$ as those of $\begin{vmatrix} j_1 & j_2 & \dots & j_m \\ A_{i_1} & i_2 & \dots & i_m \end{vmatrix}$ occupy in A . Then

$$|A| \cdot \begin{vmatrix} j_1, j_2, \dots, j_m \\ i_1, i_2, \dots, i_m \end{vmatrix} = (-1)^s |A|^m \cdot \begin{vmatrix} j_{m+1}, j_{m+2}, \dots, j_n \\ i_{m+1}, i_{m+2}, \dots, i_n \end{vmatrix}$$

where $s = i_1 + i_2 + \dots + i_m + j_1 + j_2 + \dots + j_m$.

From

$$\begin{bmatrix} a_{i_1, j_1} & a_{i_1, j_2} & \dots & a_{i_1, j_m} & a_{i_1, j_{m+1}} & \dots & a_{i_1, j_n} & \alpha_{i_1, j_1} & \alpha_{i_2, j_1} & \dots & \alpha_{i_m, j_1} & 0 & 0 & \dots & 0 \\ a_{i_2, j_1} & a_{i_2, j_2} & \dots & a_{i_2, j_m} & a_{i_2, j_{m+1}} & \dots & a_{i_2, j_n} & \alpha_{i_1, j_2} & \alpha_{i_2, j_2} & \dots & \alpha_{i_m, j_2} & 0 & 0 & \dots & 0 \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ a_{i_m, j_1} & a_{i_m, j_2} & \dots & a_{i_m, j_m} & a_{i_m, j_{m+1}} & \dots & a_{i_m, j_n} & \alpha_{i_1, j_m} & \alpha_{i_2, j_m} & \dots & \alpha_{i_m, j_m} & 0 & 0 & \dots & 0 \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ a_{i_{m+1}, j_1} & a_{i_{m+1}, j_2} & \dots & a_{i_{m+1}, j_m} & a_{i_{m+1}, j_{m+1}} & \dots & a_{i_{m+1}, j_n} & \alpha_{i_1, j_{m+1}} & \alpha_{i_2, j_{m+1}} & \dots & \alpha_{i_m, j_{m+1}} & 1 & 0 & \dots & 0 \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ a_{i_n, j_1} & a_{i_n, j_2} & \dots & a_{i_n, j_m} & a_{i_n, j_{m+1}} & \dots & a_{i_n, j_n} & \alpha_{i_1, j_n} & \alpha_{i_2, j_n} & \dots & \alpha_{i_m, j_n} & 0 & 0 & \dots & 1 \end{bmatrix}$$

$$= \begin{bmatrix} |A| & 0 & \dots & 0 & a_{i_1, j_{m+1}} & \dots & a_{i_1, j_n} \\ 0 & |A| & \dots & 0 & a_{i_2, j_{m+1}} & \dots & a_{i_2, j_n} \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ 0 & 0 & \dots & |A| & a_{i_m, j_{m+1}} & \dots & a_{i_m, j_n} \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ 0 & 0 & \dots & 0 & a_{i_{m+1}, j_{m+1}} & \dots & a_{i_{m+1}, j_n} \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots \\ 0 & 0 & \dots & 0 & a_{i_n, j_{m+1}} & \dots & a_{i_n, j_n} \end{bmatrix}$$

by taking determinants of both sides, we have

$$(-1)^s |A|^s \begin{vmatrix} j_1, j_2, \dots, j_m \\ i_1, i_2, \dots, i_m \end{vmatrix} = |A|^m \begin{vmatrix} j_{m+1}, j_{m+2}, \dots, j_n \\ i_{m+1}, i_{m+2}, \dots, i_n \end{vmatrix}$$

where s is as defined in the theorem. From this, the required form follows immediately.

7. Prove: If A is a skew-symmetric of order $2n$, then $|A|$ is the square of a polynomial in the elements of A .

By its definition, $|A|$ is a polynomial in its elements; we are to show that under the conditions given above this polynomial is a perfect square.

The theorem is true for $n=1$ since, when $A = \begin{bmatrix} 0 & a \\ -a & 0 \end{bmatrix}$, $|A| = a^2$.

Assume now that the theorem is true when $n=k$ and consider the skew-symmetric matrix $A = [a_{ij}]$ of

order $2k+2$. By partitioning, write $A = \begin{bmatrix} B & C \\ D & E \end{bmatrix}$ where $E = \begin{bmatrix} 0 & a_{2k+1, 2k+2} \\ a_{2k+2, 2k+1} & 0 \end{bmatrix}$. Then B is skew-sym-

metric of order $2k$ and, by assumption, $|B| = f^2$ where f is a polynomial in the elements of B .

If α_{ij} denotes the cofactor of a_{ij} in A , we have by Problem 6, Chapter 3, and (6.8)

$$\begin{vmatrix} \alpha_{2k+1, 2k+1} & \alpha_{2k+2, 2k+1} \\ \alpha_{2k+1, 2k+2} & \alpha_{2k+2, 2k+2} \end{vmatrix} = \begin{vmatrix} 0 & \alpha_{2k+2, 2k+1} \\ \alpha_{2k+1, 2k+2} & 0 \end{vmatrix} = |A| \cdot |B|$$

Moreover, $\alpha_{2k+2, 2k+1} = -\alpha_{2k+1, 2k+2}$; hence,

$$|A| \cdot f^2 = \alpha_{2k+1, 2k+2}^2 \quad \text{and} \quad |A| = \left\{ \frac{\alpha_{2k+1, 2k+2}}{f} \right\}^2$$

a perfect square.

SUPPLEMENTARY PROBLEMS

8. Compute the adjoint of:

$$(a) \begin{bmatrix} 1 & 2 & 3 \\ 0 & 1 & 2 \\ 0 & 0 & 0 \end{bmatrix}, \quad (b) \begin{bmatrix} 1 & 2 & 3 \\ 0 & 1 & 2 \\ 0 & 0 & 1 \end{bmatrix}, \quad (c) \begin{bmatrix} 1 & 0 & 2 \\ 2 & 1 & 0 \\ 3 & 2 & 1 \end{bmatrix}, \quad (d) \begin{bmatrix} 5 & 0 & 0 & 2 \\ 1 & 1 & 0 & 2 \\ 0 & 0 & 2 & 1 \\ 1 & 0 & 0 & 1 \end{bmatrix}$$

$$\text{Ans. } (a) \begin{bmatrix} 0 & 0 & 1 \\ 0 & 0 & -2 \\ 0 & 0 & 1 \end{bmatrix}, \quad (b) \begin{bmatrix} 1 & -2 & 1 \\ 0 & 1 & -2 \\ 0 & 0 & 1 \end{bmatrix}, \quad (c) \begin{bmatrix} 1 & 4 & -2 \\ -2 & -5 & 4 \\ 1 & -2 & 1 \end{bmatrix}, \quad (d) \begin{bmatrix} 2 & 0 & 0 & -4 \\ 2 & 6 & 0 & -16 \\ 1 & 0 & 3 & -5 \\ -2 & 0 & 0 & 10 \end{bmatrix}$$

9. Verify:

- (a) The adjoint of a scalar matrix is a scalar matrix.
- (b) The adjoint of a diagonal matrix is a diagonal matrix.
- (c) The adjoint of a triangular matrix is a triangular matrix.

10. Write a matrix $A \neq 0$ of order 3 such that $\text{adj } A = 0$.

11. If A is a 2-square matrix, show that $\text{adj}(\text{adj } A) = A$.

12. Show that the adjoint of $A = \begin{bmatrix} -1 & -2 & -2 \\ 2 & 1 & -2 \\ 2 & -2 & 1 \end{bmatrix}$ is $3A'$ and the adjoint of $A = \begin{bmatrix} -4 & -3 & -3 \\ 1 & 0 & 1 \\ 4 & 4 & 3 \end{bmatrix}$ is A itself.

13. Prove: If an n -square matrix A is of rank $< n-1$, then $\text{adj } A = 0$.

14. Prove: If A is symmetric, so also is $\text{adj } A$.

15. Prove: If A is Hermitian, so also is $\text{adj } A$.

16. Prove: If A is skew-symmetric of order n , then $\text{adj } A$ is symmetric or skew-symmetric according as n is odd or even.

17. Is there a theorem similar to that of Problem 16 for skew-Hermitian matrices?

18. For the elementary matrices, show that

$$(a) \operatorname{adj} H_{ij}^{-1} = -H_{ij}$$

$$(b) \operatorname{adj} H_i^{-1}(k) = \operatorname{diag}(1/k, 1/k, \dots, 1/k, 1, 1/k, \dots, 1/k), \text{ where the element } 1 \text{ stands in the } i\text{th row}$$

$$(c) \operatorname{adj} H_{ij}^{-1}(k) = H_{ij}(k), \text{ with similar results for the } K\text{'s.}$$

19. Prove: If A is an n -square matrix of rank n or $n-1$ and if $H_S \dots H_2 \cdot H_1 \cdot A \cdot K_1 \cdot K_2 \dots K_t = \lambda$ where λ is

$$I_n \text{ or } \begin{bmatrix} I_{n-1} & 0 \\ 0 & 0 \end{bmatrix}, \text{ then}$$

$$\operatorname{adj} A = \operatorname{adj} K_1^{-1} \cdot \operatorname{adj} K_2^{-1} \dots \operatorname{adj} K_t^{-1} \cdot \operatorname{adj} \lambda \cdot \operatorname{adj} H_S^{-1} \dots \operatorname{adj} H_2^{-1} \cdot \operatorname{adj} H_1^{-1}$$

20. Use the method of Problem 19 to compute the adjoint of

(a) A of Problem 7, Chapter 5

$$(b) \begin{bmatrix} 1 & 1 & 1 & 0 \\ 2 & 3 & 3 & 2 \\ 1 & 2 & 3 & 2 \\ 4 & 6 & 7 & 4 \end{bmatrix}$$

$$\text{Ans. (a)} \begin{bmatrix} 7 & -3 & -3 \\ -1 & 1 & 0 \\ -1 & 0 & 1 \end{bmatrix}$$

$$(b) \begin{bmatrix} -14 & 2 & -2 & 2 \\ 14 & -2 & 2 & -2 \\ 0 & 0 & 0 & 0 \\ -7 & 1 & -1 & 1 \end{bmatrix}$$

21. Let $A = [a_{ij}]$ and $B = [k - a_{ij}]$ be 3-square matrices. If $S(C) =$ sum of elements of matrix C , show that

$$S(\operatorname{adj} A) = S(\operatorname{adj} B) \quad \text{and} \quad |B| = k \cdot S(\operatorname{adj} A) - |A|$$

22. Prove: If A is n -square then $|\operatorname{adj}(\operatorname{adj} A)| = |A|^{(n-1)^2}$

23. Let $A_n = [a_{ij}]$ ($i, j = 1, 2, \dots, n$) be the lower triangular matrix whose triangle is the Pascal triangle; for example,

$$A_4 = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 \\ 1 & 2 & 1 & 0 \\ 1 & 3 & 3 & 1 \end{bmatrix}$$

Define $b_{ij} = (-1)^{i+j} a_{ij}$ and verify for $n = 2, 3, 4$ that

$$(1) \operatorname{adj} A_n = [b_{ij}] = A_n^{-1}$$

24. Let B be obtained from A by deleting its i th and p th rows and j th and q th columns. Show that

$$\begin{vmatrix} \alpha_{ij} & \alpha_{pj} \\ \alpha_{iq} & \alpha_{pq} \end{vmatrix} = (-1)^{i+p+j+q} |B| \cdot |A|$$

where α_{ij} is the cofactor of a_{ij} in $|A|$.

Chapter 7

The Inverse of a Matrix

IF A AND B are n -square matrices such that $AB = BA = I$, B is called the inverse of A , ($B = A^{-1}$) and A is called the inverse of B , ($A = B^{-1}$).

In Problem 1, we prove

I. An n -square matrix A has an inverse if and only if it is non-singular.

The inverse of a non-singular n -square matrix is unique. (See Problem 7, Chapter 2.)

II. If A is non-singular, then $AB = AC$ implies $B = C$.

THE INVERSE of a non-singular diagonal matrix $\text{diag}(k_1, k_2, \dots, k_n)$ is the diagonal matrix

$$\text{diag}(1/k_1, 1/k_2, \dots, 1/k_n)$$

If A_1, A_2, \dots, A_s are non-singular matrices, then the inverse of the direct sum $\text{diag}(A_1, A_2, \dots, A_s)$ is

$$\text{diag}(A_1^{-1}, A_2^{-1}, \dots, A_s^{-1})$$

Procedures for computing the inverse of a general non-singular matrix are given below.

INVERSE FROM THE ADJOINT. From (6.2) $A \text{adj} A = |A| \cdot I$. If A is non-singular

$$(7.1) \quad A^{-1} = \frac{\text{adj} A}{|A|} = \begin{bmatrix} \alpha_{11}/|A| & \alpha_{21}/|A| & \dots & \alpha_{n1}/|A| \\ \alpha_{12}/|A| & \alpha_{22}/|A| & \dots & \alpha_{n2}/|A| \\ \dots & \dots & \dots & \dots \\ \alpha_{1n}/|A| & \alpha_{2n}/|A| & \dots & \alpha_{nn}/|A| \end{bmatrix}$$

Example 1. From Problem 2, Chapter 6, the adjoint of $A = \begin{bmatrix} 1 & 2 & 3 \\ 1 & 3 & 4 \\ 1 & 4 & 3 \end{bmatrix}$ is $\begin{bmatrix} -7 & 6 & -1 \\ 1 & 0 & -1 \\ 1 & -2 & 1 \end{bmatrix}$.

$$\text{Since } |A| = -2, \quad A^{-1} = \frac{\text{adj} A}{|A|} = \begin{bmatrix} 7/2 & -3 & 1/2 \\ -1/2 & 0 & 1/2 \\ -1/2 & 1 & -1/2 \end{bmatrix}$$

See Problem 2.

INVERSE FROM ELEMENTARY MATRICES. Let the non-singular n -square matrix A be reduced to I by elementary transformations so that

$$H_3 \dots H_2 \cdot H_1 \cdot A \cdot K_1 \cdot K_2 \dots K_t = PAQ = I$$

Then $A = P^{-1} \cdot Q^{-1}$ by (5.5) and, since $(B^{-1})^{-1} = B$,

$$(7.2) \quad A^{-1} = (P^{-1} \cdot Q^{-1})^{-1} = Q \cdot P = K_1 \cdot K_2 \dots K_t \cdot H_3 \dots H_2 \cdot H_1$$

Example 2. From Problem 7, Chapter 5.

$$H_2 H_1 A K_1 K_2 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ -1 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ -1 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \cdot A \cdot \begin{bmatrix} 1 & -3 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & -3 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} = I$$

$$\text{Then } A^{-1} = K_1 K_2 H_2 H_1 = \begin{bmatrix} 1 & -3 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & -3 \\ 0 & 1 & 0 \\ -1 & 1 & 0 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ -1 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ -1 & 0 & 1 \end{bmatrix} = \begin{bmatrix} 7 & -3 & -3 \\ -1 & 1 & 0 \\ -1 & 0 & 1 \end{bmatrix}$$

In Chapter 5 it was shown that a non-singular matrix can be reduced to normal form by row transformations alone. Then, from (7.2) with $Q = I$, we have

$$(7.3) \quad A^{-1} = P = H_3 \dots H_2 \cdot H_1$$

That is,

III. If A is reduced to I by a sequence of row transformations alone, then A^{-1} is equal to the product in reverse order of the corresponding elementary matrices.

Example 3. Find the inverse of $A = \begin{bmatrix} 1 & 3 & 3 \\ 1 & 4 & 3 \\ 1 & 3 & 4 \end{bmatrix}$ of Example 2 using only row transformations to reduce A to I .

Write the matrix $[A \ I_3]$ and perform the sequence of row transformations which carry A into I_3 on the rows of six elements. We have

$$\begin{aligned} [A \ I_3] &= \left[\begin{array}{ccc|ccc} 1 & 3 & 3 & 1 & 0 & 0 \\ 1 & 4 & 3 & 0 & 1 & 0 \\ 1 & 3 & 4 & 0 & 0 & 1 \end{array} \right] \sim \left[\begin{array}{ccc|ccc} 1 & 3 & 3 & 1 & 0 & 0 \\ 0 & 1 & 0 & -1 & 1 & 0 \\ 0 & 0 & 1 & -1 & 0 & 1 \end{array} \right] \sim \left[\begin{array}{ccc|ccc} 1 & 0 & 3 & 4 & -3 & 0 \\ 0 & 1 & 0 & -1 & 1 & 0 \\ 0 & 0 & 1 & -1 & 0 & 1 \end{array} \right] \sim \left[\begin{array}{ccc|ccc} 1 & 0 & 0 & 7 & -3 & -3 \\ 0 & 1 & 0 & -1 & 1 & 0 \\ 0 & 0 & 1 & -1 & 0 & 1 \end{array} \right] \\ &= [I_3 \ A^{-1}] \end{aligned}$$

$$\text{by (7.3). Thus, as } A \text{ is reduced to } I_3, I_3 \text{ is carried into } A^{-1} = \begin{bmatrix} 7 & -3 & -3 \\ -1 & 1 & 0 \\ -1 & 0 & 1 \end{bmatrix}$$

See Problem 3.

INVERSE BY PARTITIONING. Let the matrix $A = [a_{ij}]$ of order n and its inverse $B = [b_{ij}]$ be partitioned into submatrices of indicated orders:

$$\begin{bmatrix} A_{11} & A_{12} \\ (p \times p) & (p \times q) \\ \hline A_{21} & A_{22} \\ (q \times p) & (q \times q) \end{bmatrix} \quad \text{and} \quad \begin{bmatrix} B_{11} & B_{12} \\ (p \times p) & (p \times q) \\ \hline B_{21} & B_{22} \\ (q \times p) & (q \times q) \end{bmatrix} \quad \text{where } p + q = n$$

Since $AB = BA = I_n$, we have

$$(7.4) \quad \begin{cases} \text{(i)} & A_{11}B_{11} + A_{12}B_{21} = I_p & \text{(iii)} & B_{21}A_{11} + B_{22}A_{21} = 0 \\ \text{(ii)} & A_{11}B_{12} + A_{12}B_{22} = 0 & \text{(iv)} & B_{21}A_{12} + B_{22}A_{22} = I_q \end{cases}$$

Then, provided A_{11} is non-singular,

$$(7.5) \quad \begin{cases} B_{11} = A_{11}^{-1} + (A_{11}^{-1} A_{12})\xi^{-1}(A_{21} A_{11}^{-1}) & B_{21} = -\xi^{-1}(A_{21} A_{11}^{-1}) \\ B_{12} = -(A_{11}^{-1} A_{12})\xi^{-1} & B_{22} = \xi^{-1} \end{cases}$$

where $\xi = A_{22} - A_{21}(A_{11}^{-1} A_{12})$.

See Problem 4.

In practice, A_{11} is usually taken of order $n-1$. To obtain A_{11}^{-1} , the following procedure is used. Let

$$G_2 = \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}, \quad G_3 = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix}, \quad G_4 = \begin{bmatrix} a_{11} & a_{12} & a_{13} & a_{14} \\ a_{21} & a_{22} & a_{23} & a_{24} \\ a_{31} & a_{32} & a_{33} & a_{34} \\ a_{41} & a_{42} & a_{43} & a_{44} \end{bmatrix}, \quad \dots$$

After computing G_2^{-1} , partition G_3 so that $A_{22} = [a_{33}]$ and use (7.5) to obtain G_3^{-1} . Repeat the process on G_4 after partitioning it so that $A_{22} = [a_{44}]$, and so on.

Example 4. Find the inverse of $A = \begin{bmatrix} 1 & 3 & 3 \\ 1 & 4 & 3 \\ 1 & 3 & 4 \end{bmatrix}$, using partitioning.

Take $A_{11} = \begin{bmatrix} 1 & 3 \\ 1 & 4 \end{bmatrix}$, $A_{12} = \begin{bmatrix} 3 \\ 3 \end{bmatrix}$, $A_{21} = [1 \ 3]$, and $A_{22} = [4]$. Now

$$A_{11}^{-1} = \begin{bmatrix} 4 & -3 \\ -1 & 1 \end{bmatrix}, \quad A_{11}^{-1}A_{12} = \begin{bmatrix} 4 & -3 \\ -1 & 1 \end{bmatrix} \begin{bmatrix} 3 \\ 3 \end{bmatrix} = \begin{bmatrix} 3 \\ 0 \end{bmatrix}, \quad A_{21}A_{11}^{-1} = [1 \ 3] \begin{bmatrix} 4 & -3 \\ -1 & 1 \end{bmatrix} = [1 \ 0].$$

$$\xi = A_{22} - A_{21}(A_{11}^{-1}A_{12}) = [4] - [1 \ 3] \begin{bmatrix} 3 \\ 0 \end{bmatrix} = [1], \quad \text{and} \quad \xi^{-1} = [1]$$

Then

$$\begin{aligned} B_{11} &= A_{11}^{-1} + (A_{11}^{-1}A_{12})\xi^{-1}(A_{21}A_{11}^{-1}) = \begin{bmatrix} 4 & -3 \\ -1 & 1 \end{bmatrix} + \begin{bmatrix} 3 \\ 0 \end{bmatrix} [1] \cdot [1 \ 0] \\ &= \begin{bmatrix} 4 & -3 \\ -1 & 1 \end{bmatrix} + \begin{bmatrix} 3 \ 0 \\ 0 \ 0 \end{bmatrix} = \begin{bmatrix} 7 & -3 \\ -1 & 1 \end{bmatrix} \end{aligned}$$

$$B_{12} = -(A_{11}^{-1}A_{12})\xi^{-1} = \begin{bmatrix} -3 \\ 0 \end{bmatrix}$$

$$B_{21} = -\xi^{-1}(A_{21}A_{11}^{-1}) = [-1 \ 0]$$

$$B_{22} = \xi^{-1} = [1]$$

and

$$A^{-1} = \begin{bmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{bmatrix} = \begin{bmatrix} 7 & -3 & -3 \\ -1 & 1 & 0 \\ -1 & 0 & 1 \end{bmatrix}$$

See Problems 5-6.

THE INVERSE OF A SYMMETRIC MATRIX. When A is symmetric, $\alpha_{ij} = \alpha_{ji}$ and only $\frac{1}{2}n(n+1)$ cofactors need be computed instead of the usual n^2 in obtaining A^{-1} from $\text{adj} A$.

If there is to be any gain in computing A^{-1} as the product of elementary matrices, the elementary transformations must be performed so that the property of being symmetric is preserved. This requires that the transformations occur in pairs, a row transformation followed immediately by the same column transformation. For example,

$$H_{12} \begin{bmatrix} 0 & b & c & \dots \\ b & a & \dots & \dots \\ c & \dots & \dots & \dots \\ \vdots & & & \end{bmatrix} K_{12} = \begin{bmatrix} a & b & \dots & \dots \\ b & 0 & c & \dots \\ \vdots & c & \dots & \dots \\ \vdots & & & \end{bmatrix}$$

$$H_{21} \left(-\frac{b}{a}\right) \begin{bmatrix} a & b & c & \dots \\ b & & & \dots \\ c & & & \dots \\ \vdots & & & \end{bmatrix} K_{21} \left(-\frac{b}{a}\right) = \begin{bmatrix} a & 0 & c & \dots \\ 0 & & & \dots \\ c & & & \dots \\ \vdots & & & \end{bmatrix}$$

However, when the element a in the diagonal is replaced by 1, the pair of transformations are $H_1(1/\sqrt{a})$ and $K_1(1/\sqrt{a})$. In general, \sqrt{a} is either irrational or imaginary; hence, this procedure is not recommended.

The maximum gain occurs when the method of partitioning is used since then (7.5) reduces to

$$(7.6) \quad \begin{aligned} B_{11} &= A_{11}^{-1} + (A_{11}^{-1}A_{12})\xi^{-1}(A_{11}^{-1}A_{12})', & B_{21} &= B'_{12} \\ B_{12} &= -(A_{11}^{-1}A_{12})\xi^{-1}, & B_{22} &= \xi^{-1} \end{aligned}$$

where $\xi = A_{22} - A_{21}(A_{11}^{-1}A_{12})$.

See Problem 7.

When A is not symmetric, the above procedure may be used to find the inverse of AA' , which is symmetric, and then the inverse of A is found by

$$(7.7) \quad A^{-1} = (AA')^{-1}A'$$

SOLVED PROBLEMS

1. Prove: An n -square matrix A has an inverse if and only if it is non-singular.

Suppose A is non-singular. By Theorem IV, Chapter 5, there exist non-singular matrices P and Q such that $PAQ = I$. Then $A = P^{-1} \cdot Q^{-1}$ and $A^{-1} = Q \cdot P$ exists.

Suppose A^{-1} exists. The $A \cdot A^{-1} = I$ is of rank n . If A were singular, AA^{-1} would be of rank $< n$; hence, A is non-singular.

2. (a) When $A = \begin{bmatrix} 2 & 3 \\ 1 & 4 \end{bmatrix}$, then $|A| = 5$, $\text{adj } A = \begin{bmatrix} 4 & -3 \\ -1 & 2 \end{bmatrix}$, and $A^{-1} = \begin{bmatrix} 4/5 & -3/5 \\ -1/5 & 2/5 \end{bmatrix}$.

(b) When $A = \begin{bmatrix} 2 & 3 & 1 \\ 1 & 2 & 3 \\ 3 & 1 & 2 \end{bmatrix}$, then $|A| = 18$, $\text{adj } A = \begin{bmatrix} 1 & -5 & 7 \\ 7 & 1 & -5 \\ -5 & 7 & 1 \end{bmatrix}$, and $A^{-1} = \frac{1}{18} \begin{bmatrix} 1 & -5 & 7 \\ 7 & 1 & -5 \\ -5 & 7 & 1 \end{bmatrix}$.

3. Find the inverse of $A = \begin{bmatrix} 2 & 4 & 3 & 2 \\ 3 & 6 & 5 & 2 \\ 2 & 5 & 2 & -3 \\ 4 & 5 & 14 & 14 \end{bmatrix}$

$$[A \ I_4] = \left[\begin{array}{cccc|cccc} 2 & 4 & 3 & 2 & 1 & 0 & 0 & 0 \\ 3 & 6 & 5 & 2 & 0 & 1 & 0 & 0 \\ 2 & 5 & 2 & -3 & 0 & 0 & 1 & 0 \\ 4 & 5 & 14 & 14 & 0 & 0 & 0 & 1 \end{array} \right] \sim \left[\begin{array}{cccc|cccc} 1 & 2 & 3/2 & 1 & 1/2 & 0 & 0 & 0 \\ 3 & 6 & 5 & 2 & 0 & 1 & 0 & 0 \\ 2 & 5 & 2 & -3 & 0 & 0 & 1 & 0 \\ 4 & 5 & 14 & 14 & 0 & 0 & 0 & 1 \end{array} \right] \sim \left[\begin{array}{cccc|cccc} 1 & 2 & 3/2 & 1 & 1/2 & 0 & 0 & 0 \\ 0 & 0 & 1/2 & -1 & -3/2 & 1 & 0 & 0 \\ 0 & 1 & -1 & -5 & -1 & 0 & 1 & 0 \\ 0 & -3 & 8 & 10 & -2 & 0 & 0 & 1 \end{array} \right]$$

$$\sim \left[\begin{array}{cccc|cccc} 1 & 2 & 3/2 & 1 & 1/2 & 0 & 0 & 0 \\ 0 & 1 & -1 & -5 & -1 & 0 & 1 & 0 \\ 0 & 0 & 1/2 & -1 & -3/2 & 1 & 0 & 0 \\ 0 & -3 & 8 & 10 & -2 & 0 & 0 & 1 \end{array} \right] \sim \left[\begin{array}{cccc|cccc} 1 & 0 & 7/2 & 11 & 5/2 & 0 & -2 & 0 \\ 0 & 1 & -1 & -5 & -1 & 0 & 1 & 0 \\ 0 & 0 & 1 & -2 & -3 & 2 & 0 & 0 \\ 0 & 0 & 5 & -5 & -5 & 0 & 3 & 1 \end{array} \right]$$

$$\sim \left[\begin{array}{cccc|cccc} 1 & 0 & 0 & 18 & 13 & -7 & -2 & 0 \\ 0 & 1 & 0 & -7 & -4 & 2 & 1 & 0 \\ 0 & 0 & 1 & -2 & -3 & 2 & 0 & 0 \\ 0 & 0 & 0 & 5 & 10 & -10 & 3 & 1 \end{array} \right] \sim \left[\begin{array}{cccc|cccc} 1 & 0 & 0 & 18 & 13 & -7 & -2 & 0 \\ 0 & 1 & 0 & -7 & -4 & 2 & 1 & 0 \\ 0 & 0 & 1 & -2 & -3 & 2 & 0 & 0 \\ 0 & 0 & 0 & 1 & 2 & -2 & 3/5 & 1/5 \end{array} \right]$$

$$\sim \left[\begin{array}{cccc|cccc} 1 & 0 & 0 & 0 & -23 & 29 & -64/5 & -18/5 \\ 0 & 1 & 0 & 0 & 10 & -12 & 26/5 & 7/5 \\ 0 & 0 & 1 & 0 & 1 & -2 & 6/5 & 2/5 \\ 0 & 0 & 0 & 1 & 2 & -2 & 3/5 & 1/5 \end{array} \right]$$

= $[I_4 \ A^{-1}]$

The inverse is $A^{-1} = \begin{bmatrix} -23 & 29 & -64/5 & -18/5 \\ 10 & -12 & 26/5 & 7/5 \\ 1 & -2 & 6/5 & 2/5 \\ 2 & -2 & 3/5 & 1/5 \end{bmatrix}$

4. Solve $\begin{cases} \text{(i)} & A_{11}B_{11} + A_{12}B_{21} = I & \text{(iii)} & B_{21}A_{11} + B_{22}A_{21} = 0 \\ \text{(ii)} & A_{11}B_{12} + A_{12}B_{22} = 0 & \text{(iv)} & B_{21}A_{12} + B_{22}A_{22} = I \end{cases}$ for B_{11}, B_{12}, B_{21} , and B_{22} .

Set $B_{22} = \xi^{-1}$. From (ii), $B_{12} = -(A_{11}^{-1}A_{12})\xi^{-1}$; from (iii), $B_{21} = -\xi^{-1}(A_{21}A_{11}^{-1})$; and, from (i), $B_{11} = A_{11}^{-1} - A_{11}^{-1}A_{12}B_{21} = A_{11}^{-1} + (A_{11}^{-1}A_{12})\xi^{-1}(A_{21}A_{11}^{-1})$.

Finally, substituting in (iv).

$$-\xi^{-1}(A_{21}A_{11}^{-1})A_{12} + \xi^{-1}A_{22} = I \quad \text{and} \quad \xi = A_{22} - (A_{21}A_{11}^{-1})A_{12}$$

5. Find the inverse of $A = \begin{bmatrix} 1 & 2 & 3 & 1 \\ 1 & 3 & 3 & 2 \\ 2 & 4 & 3 & 3 \\ 1 & 1 & 1 & 1 \end{bmatrix}$ by partitioning.

(a) Take $G_3 = \begin{bmatrix} 1 & 2 & 3 \\ 1 & 3 & 3 \\ 2 & 4 & 3 \end{bmatrix}$ and partition so that

$$A_{11} = \begin{bmatrix} 1 & 2 \\ 1 & 3 \end{bmatrix}, \quad A_{12} = \begin{bmatrix} 3 \\ 3 \end{bmatrix}, \quad A_{21} = [2 \ 4], \quad \text{and} \quad A_{22} = [3]$$

$$\text{Now } A_{11}^{-1} = \begin{bmatrix} 3 & -2 \\ -1 & 1 \end{bmatrix}, \quad A_{11}^{-1}A_{12} = \begin{bmatrix} 3 & -2 \\ -1 & 1 \end{bmatrix} \begin{bmatrix} 3 \\ 3 \end{bmatrix} = \begin{bmatrix} 3 \\ 0 \end{bmatrix}, \quad A_{21}A_{11}^{-1} = [2 \ 4] \begin{bmatrix} 3 & -2 \\ -1 & 1 \end{bmatrix} = [2 \ 0].$$

$$\xi = A_{22} - A_{21}(A_{11}^{-1}A_{12}) = [3] - [2 \ 4] \begin{bmatrix} 3 \\ 0 \end{bmatrix} = [-3], \quad \text{and} \quad \xi^{-1} = [-1/3]$$

$$\begin{aligned} \text{Then } B_{11} &= A_{11}^{-1} + (A_{11}^{-1}A_{12})\xi^{-1}(A_{21}A_{11}^{-1}) = \begin{bmatrix} 3 & -2 \\ -1 & 1 \end{bmatrix} + \begin{bmatrix} 3 \\ 0 \end{bmatrix} \begin{bmatrix} -1 \\ 3 \end{bmatrix} [2 \ 0] = \begin{bmatrix} 3 & -2 \\ -1 & 1 \end{bmatrix} - \begin{bmatrix} 2 & 0 \\ 0 & 0 \end{bmatrix} \\ &= \frac{1}{3} \begin{bmatrix} 3 & -6 \\ -3 & 3 \end{bmatrix} \end{aligned}$$

$$B_{12} = -(A_{11}^{-1}A_{12})\xi^{-1} = \frac{1}{3} \begin{bmatrix} 3 \\ 0 \end{bmatrix}, \quad B_{21} = -\xi^{-1}(A_{21}A_{11}^{-1}) = \frac{1}{3}[2 \ 0], \quad B_{22} = \xi^{-1} = \begin{bmatrix} -1/3 \end{bmatrix}$$

$$\text{and } G_3^{-1} = \begin{bmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{bmatrix} = \frac{1}{3} \begin{bmatrix} 3 & -6 & 3 \\ -3 & 3 & 0 \\ 2 & 0 & -1 \end{bmatrix}$$

(b) Partition A so that $A_{11} = \begin{bmatrix} 1 & 2 & 3 \\ 1 & 3 & 3 \\ 2 & 4 & 3 \end{bmatrix}$, $A_{12} = \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$, $A_{21} = [1 \ 1 \ 1]$, and $A_{22} = [1]$.

$$\text{Now } A_{11}^{-1} = \frac{1}{3} \begin{bmatrix} -3 & -6 & 3 \\ -3 & 3 & 0 \\ 2 & 0 & -1 \end{bmatrix}, \quad A_{11}^{-1}A_{12} = \frac{1}{3} \begin{bmatrix} 0 \\ 3 \\ -1 \end{bmatrix}, \quad A_{21}A_{11}^{-1} = \frac{1}{3}[2 \ -3 \ 2].$$

$$\xi = [1] - [1 \ 1 \ 1] \left(\frac{1}{3}\right) \begin{bmatrix} 0 \\ 3 \\ -1 \end{bmatrix} = \begin{bmatrix} 1/3 \end{bmatrix}, \quad \text{and} \quad \xi^{-1} = [3]$$

$$\text{Then } B_{11} = \frac{1}{3} \begin{bmatrix} 3 & -6 & 3 \\ -3 & 3 & 0 \\ 2 & 0 & -1 \end{bmatrix} + \frac{1}{3} \begin{bmatrix} 0 \\ 3 \\ -1 \end{bmatrix} [3] \frac{1}{3} [2 \ -3 \ 2] = \frac{1}{3} \begin{bmatrix} 3 & -6 & 3 \\ -3 & 3 & 0 \\ 2 & 0 & -1 \end{bmatrix} + \frac{1}{3} \begin{bmatrix} 0 & 0 & 0 \\ 6 & -9 & 6 \\ -2 & 3 & -2 \end{bmatrix} = \begin{bmatrix} 1 & -2 & 1 \\ 1 & -2 & 2 \\ 0 & 1 & -1 \end{bmatrix}$$

$$B_{12} = \begin{bmatrix} 0 \\ -3 \\ 1 \end{bmatrix}, \quad B_{21} = [-2 \ 3 \ -2], \quad B_{22} = [3]$$

$$\text{and } A^{-1} = \begin{bmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{bmatrix} = \begin{bmatrix} 1 & -2 & 1 & 0 \\ 1 & -2 & 2 & -3 \\ 0 & 1 & -1 & 1 \\ -2 & 3 & -2 & 3 \end{bmatrix}$$

6. Find the inverse of $A = \begin{bmatrix} 1 & 3 & 3 \\ 1 & 3 & 4 \\ 1 & 4 & 3 \end{bmatrix}$ by partitioning.

We cannot take $A_{11} = \begin{bmatrix} 1 & 3 \\ 1 & 3 \end{bmatrix}$ since this is singular.

By Example 3, the inverse of $H_{23}A = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix} A = \begin{bmatrix} 1 & 3 & 3 \\ 1 & 4 & 3 \\ 1 & 3 & 4 \end{bmatrix} = B$ is $B^{-1} = \begin{bmatrix} 7 & -3 & -3 \\ -1 & 1 & 0 \\ -1 & 0 & 1 \end{bmatrix}$. Then

$$A^{-1} = B^{-1}H_{23} = \begin{bmatrix} 7 & -3 & -3 \\ -1 & 1 & 0 \\ -1 & 0 & 1 \end{bmatrix} \cdot \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix} = \begin{bmatrix} 7 & -3 & -3 \\ -1 & 0 & 1 \\ -1 & 1 & 0 \end{bmatrix}$$

Thus, if the $(n-1)$ -square minor A_{11} of the n -square non-singular matrix A is singular, we first bring a non-singular $(n-1)$ -square matrix into the upper left corner to obtain B , find the inverse of B , and by the proper transformation on B^{-1} obtain A^{-1} .

7. Compute the inverse of the symmetric matrix $A = \begin{bmatrix} 2 & 1 & -1 & 2 \\ 1 & 3 & 2 & -3 \\ -1 & 2 & 1 & -1 \\ 2 & -3 & -1 & 4 \end{bmatrix}$

Consider first the submatrix $G_3 = \begin{bmatrix} 2 & 1 & -1 \\ 1 & 3 & 2 \\ -1 & 2 & 1 \end{bmatrix}$ partitioned so that

$$A_{11} = \begin{bmatrix} 2 & 1 \\ 1 & 3 \end{bmatrix}, \quad A_{12} = \begin{bmatrix} -1 \\ 2 \end{bmatrix}, \quad A_{21} = [-1 \ 2], \quad A_{22} = [1]$$

Now $A_{11}^{-1} = \begin{bmatrix} 3/5 & -1/5 \\ -1/5 & 2/5 \end{bmatrix}, \quad A_{11}^{-1}A_{12} = \begin{bmatrix} -3/5 & -1/5 \\ -1/5 & 2/5 \end{bmatrix} \begin{bmatrix} -1 \\ 2 \end{bmatrix} = \begin{bmatrix} -1 \\ 1 \end{bmatrix}$

$$\xi = A_{22} - A_{21}(A_{11}^{-1}A_{12}) = [1] - [-1 \ 2] \begin{bmatrix} -1 \\ 1 \end{bmatrix} = [-2] \quad \text{and} \quad \xi^{-1} = [-\frac{1}{2}]$$

Then $B_{11} = \begin{bmatrix} 3/5 & -1/5 \\ -1/5 & 2/5 \end{bmatrix} + \begin{bmatrix} -1 \\ 1 \end{bmatrix} [-\frac{1}{2}] [-1 \ 1] = \begin{bmatrix} 3/5 & -1/5 \\ -1/5 & 2/5 \end{bmatrix} + \begin{bmatrix} -\frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & -\frac{1}{2} \end{bmatrix} = \frac{1}{10} \begin{bmatrix} 1 & 3 \\ 3 & -1 \end{bmatrix}$

$$B_{12} = \begin{bmatrix} -\frac{1}{2} \\ \frac{1}{2} \end{bmatrix}, \quad B_{21} = [-\frac{1}{2} \ \frac{1}{2}], \quad B_{22} = [-\frac{1}{2}]$$

and

$$G_3^{-1} = \frac{1}{10} \begin{bmatrix} 1 & 3 & -5 \\ 3 & -1 & 5 \\ -5 & 5 & -5 \end{bmatrix}$$

Consider now the matrix A partitioned so that

$$A_{11} = \begin{bmatrix} 2 & 1 & -1 \\ 1 & 3 & 2 \\ -1 & 2 & 1 \end{bmatrix}, \quad A_{12} = \begin{bmatrix} 2 \\ -3 \\ -1 \end{bmatrix}, \quad A_{21} = [2 \ -3 \ -1], \quad A_{22} = [4]$$

Now $A_{11}^{-1} = \frac{1}{10} \begin{bmatrix} 1 & 3 & -5 \\ -3 & -1 & 5 \\ -5 & 5 & -5 \end{bmatrix}, \quad A_{11}^{-1}A_{12} = \begin{bmatrix} -1/5 \\ 2/5 \\ -2 \end{bmatrix}, \quad \xi = [18/5], \quad \xi^{-1} = [5/18].$

$$\text{Then } B_{11} = \frac{1}{18} \begin{bmatrix} 2 & 5 & -7 \\ 5 & -1 & 5 \\ -7 & 5 & 11 \end{bmatrix}, \quad B_{12} = \frac{1}{18} \begin{bmatrix} 1 \\ -2 \\ 10 \end{bmatrix}, \quad B_{21} = \frac{1}{18} [1 \ -2 \ 10], \quad B_{22} = [5/18]$$

and

$$A^{-1} = \frac{1}{18} \begin{bmatrix} 2 & 5 & -7 & 1 \\ 5 & -1 & 5 & -2 \\ -7 & 5 & 11 & 10 \\ 1 & -2 & 10 & 5 \end{bmatrix}$$

SUPPLEMENTARY PROBLEMS

8. Find the adjoint and inverse of each of the following:

$$(a) \begin{bmatrix} 1 & 2 & -1 \\ -1 & 1 & 2 \\ 2 & -1 & 1 \end{bmatrix}$$

$$(b) \begin{bmatrix} 2 & 3 & 4 \\ 4 & 3 & 1 \\ 1 & 2 & 4 \end{bmatrix}$$

$$(c) \begin{bmatrix} 1 & 2 & 3 \\ 2 & 4 & 5 \\ 3 & 5 & 6 \end{bmatrix}$$

$$(d) \begin{bmatrix} 1 & 2 & 0 & 0 \\ 0 & 3 & 0 & 0 \\ 0 & 0 & 2 & 1 \\ 0 & 0 & 0 & 3 \end{bmatrix}$$

$$\text{Ans. Inverses } (a) \frac{1}{14} \begin{bmatrix} 3 & -1 & 5 \\ 5 & 3 & -1 \\ -1 & 5 & 3 \end{bmatrix}, \quad (b) \frac{1}{5} \begin{bmatrix} -10 & 4 & 9 \\ 15 & -4 & -14 \\ -5 & 1 & 6 \end{bmatrix}, \quad (c) \begin{bmatrix} 1 & -3 & 2 \\ -3 & 3 & -1 \\ 2 & -1 & 0 \end{bmatrix}, \quad (d) \begin{bmatrix} 1 & -2/3 & 0 & 0 \\ 0 & 1/3 & 0 & 0 \\ 0 & 0 & 1/2 & -1/6 \\ 0 & 0 & 0 & 1/3 \end{bmatrix}$$

9. Find the inverse of the matrix of Problem 8(d) as a direct sum.

10. Obtain the inverses of the matrices of Problem 8 using the method of Problem 3.

$$11. \text{ Same, for the matrices } (a) \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 2 & 3 & -4 \\ 2 & 3 & 5 & -5 \\ 3 & -4 & -5 & 8 \end{bmatrix}, \quad (b) \begin{bmatrix} 3 & 4 & 2 & 7 \\ 2 & 3 & 3 & 2 \\ 5 & 7 & 3 & 9 \\ 2 & 3 & 2 & 3 \end{bmatrix}, \quad (c) \begin{bmatrix} 2 & 5 & 2 & 3 \\ 2 & 3 & 3 & 4 \\ 3 & 6 & 3 & 2 \\ 4 & 12 & 0 & 8 \end{bmatrix}, \quad (d) \begin{bmatrix} 1 & 3 & 3 & 2 & -1 \\ 1 & 4 & 3 & 3 & -1 \\ 1 & 3 & 4 & 1 & 1 \\ 1 & 1 & 1 & 1 & -1 \\ 1 & -2 & -1 & 2 & 2 \end{bmatrix}$$

$$\text{Ans. } (a) \frac{1}{18} \begin{bmatrix} 2 & 16 & -6 & 4 \\ 22 & 41 & -30 & -1 \\ -10 & -44 & 30 & -2 \\ 4 & -13 & 6 & -1 \end{bmatrix}$$

$$(c) \frac{1}{48} \begin{bmatrix} -144 & 36 & 60 & 21 \\ 48 & -20 & -12 & -5 \\ 48 & -4 & -12 & -13 \\ 0 & 12 & -12 & 3 \end{bmatrix}$$

$$(b) \frac{1}{2} \begin{bmatrix} -1 & 11 & 7 & -26 \\ -1 & -7 & -3 & 16 \\ 1 & 1 & -1 & 0 \\ 1 & -1 & -1 & 2 \end{bmatrix}$$

$$(d) \frac{1}{15} \begin{bmatrix} 30 & -20 & -15 & 25 & -5 \\ 30 & -11 & -18 & 7 & -8 \\ -30 & 12 & 21 & -9 & 6 \\ -15 & 12 & 6 & -9 & 6 \\ 15 & -7 & -6 & -1 & -1 \end{bmatrix}$$

12. Use the result of Example 4 to obtain the inverse of the matrix of Problem 11(d) by partitioning.

13. Obtain by partitioning the inverses of the matrices of Problems 8(a), 8(b), 11(a)–11(c).

14. Obtain by partitioning the inverses of the symmetric matrices (a) $\begin{bmatrix} 1 & 2 & -1 & 2 \\ 2 & 2 & -1 & 1 \\ -1 & -1 & 1 & -1 \\ 2 & 1 & -1 & 2 \end{bmatrix}$, (b) $\begin{bmatrix} 0 & 1 & 2 & 2 \\ 1 & 1 & 2 & 3 \\ 2 & 2 & 2 & 3 \\ 2 & 3 & 3 & 3 \end{bmatrix}$

Ans. (a) $-\frac{1}{2} \begin{bmatrix} 1 & -1 & -1 & -1 \\ -1 & -1 & -1 & 1 \\ -1 & -1 & -5 & -1 \\ -1 & 1 & -1 & -1 \end{bmatrix}$, (b) $\begin{bmatrix} -3 & 3 & -3 & 2 \\ 3 & -4 & 4 & -2 \\ -3 & 4 & -5 & 3 \\ 2 & -2 & 3 & -2 \end{bmatrix}$

- 15. Prove: If A is non-singular, then $AB = AC$ implies $B = C$.
- 16. Show that if the non-singular matrices A and B commute, so also do
 (a) A^{-1} and B , (b) A and B^{-1} , (c) A^{-1} and B^{-1} . Hint: (a) $A^{-1}(AB)A^{-1} = A^{-1}(BA)A^{-1}$.
- 17. Show that if the non-singular matrix A is symmetric, so also is A^{-1} .
 Hint: $A^{-1}A = I = (AA^{-1})' = (A^{-1})'A$.
- 18. Show that if the non-singular symmetric matrices A and B commute, then (a) $A^{-1}B$, (b) AB^{-1} , and (c) $A^{-1}B^{-1}$ are symmetric. Hint: (a) $(A^{-1}B)' = (BA^{-1})' = (A^{-1})'B' = A^{-1}B$.
- 19. An $m \times n$ matrix A is said to have a *right inverse* B if $AB = I$ and a *left inverse* C if $CA = I$. Show that A has a right inverse if and only if A is of rank m and has a left inverse if and only if the rank of A is n .

20. Find a right inverse of $A = \begin{bmatrix} 1 & 3 & 2 & 3 \\ 1 & 4 & 1 & 3 \\ 1 & 3 & 5 & 4 \end{bmatrix}$ if one exists.

Hint. The rank of A is 3 and the submatrix $S = \begin{bmatrix} 1 & 3 & 2 \\ 1 & 4 & 1 \\ 1 & 3 & 5 \end{bmatrix}$ is non-singular with inverse S^{-1} . A right inverse of

A is the 4×3 matrix $B = \begin{bmatrix} S^{-1} \\ 0 \end{bmatrix} = \frac{1}{3} \begin{bmatrix} 17 & -9 & -5 \\ -4 & 3 & 1 \\ -1 & 0 & 1 \\ 0 & 0 & 0 \end{bmatrix}$

21. Show that the submatrix $T = \begin{bmatrix} 1 & 3 & 3 \\ 1 & 4 & 3 \\ 1 & 3 & 4 \end{bmatrix}$ of A of Problem 20 is non-singular and obtain $\begin{bmatrix} 7 & -3 & -3 \\ -1 & 1 & 0 \\ 0 & 0 & 0 \\ -1 & 0 & 1 \end{bmatrix}$ as another right inverse of A .

22. Obtain $\begin{bmatrix} 7 & -1 & -1 & a \\ -3 & 1 & 0 & b \\ -3 & 0 & 1 & c \end{bmatrix}$ as a left inverse of $\begin{bmatrix} 1 & 1 & 1 \\ 3 & 4 & 3 \\ 3 & 3 & 4 \\ 0 & 0 & 0 \end{bmatrix}$ where a , b , and c are arbitrary.

23. Show that $A = \begin{bmatrix} 1 & 3 & 4 & 7 \\ 1 & 4 & 5 & 9 \\ 2 & 3 & 5 & 8 \end{bmatrix}$ has neither a right nor a left inverse.

24. Prove: If $|A_{11}| \neq 0$, then $\begin{vmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{vmatrix} = |A_{11}| \cdot |A_{22} - A_{21}A_{11}^{-1}A_{12}|$.

25. If $|I + A| \neq 0$, then $(I + A)^{-1}$ and $(I - A)$ commute.

26. Prove: (i) of Problem 23, Chapter 6.

Chapter 8

Fields

NUMBER FIELDS. A collection or set S of real or complex numbers, consisting of more than the element 0, is called a **number field** provided the operations of addition, subtraction, multiplication, and division (except by 0) on any two of the numbers yield a number of S .

Examples of number fields are:

- (a) the set of all rational numbers,
- (b) the set of all real numbers,
- (c) the set of all numbers of the form $a + b\sqrt{3}$, where a and b are rational numbers,
- (d) the set of all complex numbers $a + bi$, where a and b are real numbers.

The set of all integers and the set of all numbers of the form $b\sqrt{3}$, where b is a rational number, are **not** number fields.

GENERAL FIELDS. A collection or set S of two or more elements, together with two operations called addition (+) and multiplication (\cdot), is called a **field** F provided that (a, b, c, \dots are elements of F , i.e. scalars),

$$A_1: a + b \text{ is a unique element of } F$$

$$A_2: a + b = b + a$$

$$A_3: a + (b + c) = (a + b) + c$$

$$A_4: \text{For every element } a \text{ in } F \text{ there exists an element } 0 \text{ in } F \text{ such that } a + 0 = 0 + a = a.$$

$$A_5: \text{For each element } a \text{ in } F \text{ there exists a unique element } -a \text{ in } F \text{ such that } a + (-a) = 0.$$

$$M_1: ab = a \cdot b \text{ is a unique element of } F$$

$$M_2: ab = ba$$

$$M_3: (ab)c = a(bc)$$

$$M_4: \text{For every element } a \text{ in } F \text{ there exists an element } 1 \neq 0 \text{ such that } 1 \cdot a = a \cdot 1 = a.$$

$$M_5: \text{For each element } a \neq 0 \text{ in } F \text{ there exists a unique element } a^{-1} \text{ in } F \text{ such that } a \cdot a^{-1} = a^{-1} \cdot a = 1.$$

$$D_1: a(b + c) = ab + ac$$

$$D_2: (a + b)c = ac + bc$$

In addition to the number fields listed above, other examples of fields are:

(e) the set of all quotients $\frac{P(x)}{Q(x)}$ of polynomials in x with real coefficients,

(f) the set of all 2×2 matrices of the form $\begin{bmatrix} a & -b \\ b & a \end{bmatrix}$ where a and b are real numbers.

(g) the set in which $a + a = 0$. This field, called of **characteristic 2**, will be excluded hereafter. In this field, for example, the customary proof that a determinant having two rows identical is 0 is not valid. By interchanging the two identical rows, we are led to $D = -D$ or $2D = 0$; but D is not necessarily 0.

SUBFIELDS. If S and T are two sets and if every member of S is also a member of T , then S is called a subset of T .

If S and T are fields and if S is a subset of T , then S is called a **subfield** of T . For example, the field of all real numbers is a subfield of the field of all complex numbers; the field of all rational numbers is a subfield of the field of all real numbers and the field of all complex numbers.

MATRICES OVER A FIELD. When all of the elements of a matrix A are in a field F , we say that " A is over F ". For example,

$$A = \begin{bmatrix} 1 & 1/2 \\ 1/4 & 2/3 \end{bmatrix} \text{ is over the rational field and } B = \begin{bmatrix} 1 & 1+i \\ 2 & 1-3i \end{bmatrix} \text{ is over the complex field}$$

Here, A is also over the real field while B is not; also A is over the complex field.

Let A, B, C, \dots be matrices over the same field F and let F be the smallest field which contains all the elements; that is, if all the elements are rational numbers, the field F is the rational field and not the real or complex field. An examination of the various operations defined on these matrices, individually or collectively, in the previous chapters shows that no elements other than those in F are ever required. For example:

The sum, difference, and product are matrices over F .

If A is non-singular, its inverse is over F .

If $A \sim I$ then there exist matrices P and Q over F such that $PAQ = I$ and I is over F .

If A is over the rational field and is of rank r , its rank is unchanged when considered over the real or the complex field.

Hereafter when A is said to be over F it will be assumed that F is the smallest field containing all of its elements.

In later chapters it will at times be necessary to restrict the field, say, to the real field. At other times, the field of the elements will be extended, say, from the rational field to the real field. Otherwise, the statement " A over F " implies no restriction on the field, **except for the excluded field of characteristic two.**

SOLVED PROBLEM

1. Verify that the set of all complex numbers constitutes a field.

To do this we simply check the properties A_1-A_5, M_1-M_5 , and D_1-D_2 . The zero element (A_4) is 0 and the unit element (M_4) is 1. If $a+bi$ and $c+di$ are two elements, the negative (A_5) of $a+bi$ is $-a-bi$, the product (M_1) is $(a+bi)(c+di) = (ac-bd) + (ad+bc)i$; the inverse (M_5) of $a+bi \neq 0$ is

$$\frac{1}{a+bi} = \frac{a-bi}{a^2+b^2} = \frac{a}{a^2+b^2} - \frac{bi}{a^2+b^2}$$

Verification of the remaining properties is left as an exercise for the reader.

SUPPLEMENTARY PROBLEMS

2. Verify (a) the set of all real numbers of the form $a + b\sqrt{5}$ where a and b are rational numbers and
 (b) the set of all quotients $\frac{P(x)}{Q(x)}$ of polynomials in x with real coefficients constitute fields.
3. Verify (a) the set of all rational numbers.
 (b) the set of all numbers $a + b\sqrt{3}$, where a and b are rational numbers, and
 (c) the set of all numbers $a + bi$, where a and b are rational numbers are subfields of the complex field.
4. Verify that the set of all 2×2 matrices of the form $\begin{bmatrix} a & -b \\ b & a \end{bmatrix}$, where a and b are rational numbers, forms a field.
 Show that this is a subfield of the field of all 2×2 matrices of the form $\begin{bmatrix} a & -b \\ b & a \end{bmatrix}$ where a and b are real numbers.
5. Why does not the set of all 2×2 matrices with real elements form a field?
6. A set R of elements a, b, c, \dots satisfying the conditions ($A_1, A_2, A_3, A_4, A_5; M_1, M_3, D_1, D_2$) of Page 64 is called a **ring**. To emphasize the fact that multiplication is not commutative, R may be called a **non-commutative ring**. When a ring R satisfies M_2 , it is called **commutative**. When a ring R satisfies M_4 , it is spoken of as a **ring with unit element**.
 Verify:
 (a) the set of even integers $0, \pm 2, \pm 4, \dots$ is an example of a commutative ring without unit element.
 (b) the set of all integers $0, \pm 1, \pm 2, \pm 3, \dots$ is an example of a commutative ring with unit element.
 (c) the set of all n -square matrices over F is an example of a non-commutative ring with unit element.
 (d) the set of all 2×2 matrices of the form $\begin{bmatrix} a & -b \\ b & a \end{bmatrix}$, where a and b are real numbers, is an example of a commutative ring with unit element.
7. Can the set (a) of Problem 6 be turned into a commutative ring with unit element by simply adjoining the elements ± 1 to the set?
8. By Problem 4, the set (d) of Problem 6 is a field. Is every field a ring? Is every commutative ring with unit element a field?
9. Describe the ring of all 2×2 matrices $\begin{bmatrix} 0 & 0 \\ a & b \end{bmatrix}$, where a and b are in F . If A is any matrix of the ring and $L = \begin{bmatrix} 0 & 0 \\ 1 & 1 \end{bmatrix}$, show that $LA = A$. Call L a **left unit element**. Is there a right unit element?
10. Let C be the field of all complex numbers $p + qi$ and K be the field of all 2×2 matrices $\begin{bmatrix} u & -v \\ v & u \end{bmatrix}$ where p, q, u, v are real numbers. Take the complex number $a + bi$ and the matrix $\begin{bmatrix} a & -b \\ b & a \end{bmatrix}$ as corresponding elements of the two sets and call each the **image** of the other.
 (a) Write the image of $\begin{bmatrix} 2 & -3 \\ 3 & 2 \end{bmatrix}$, $\begin{bmatrix} 0 & -4 \\ 4 & 0 \end{bmatrix}$; $3 + 2i$; 5.
 (b) Show that the image of the sum (product) of two elements of K is the sum (product) of their images in C .
 (c) Show that the image of the identity element of K is the identity element of C .
 (d) What is the image of the conjugate of $a + bi$?
 (e) What is the image of the inverse of $\begin{bmatrix} a & -b \\ b & a \end{bmatrix}$?

This is an example of an **isomorphism** between two sets.

Chapter 9

Linear Dependence of Vectors and Forms

THE ORDERED PAIR of real numbers (x_1, x_2) is used to denote a point X in a plane. The same pair of numbers, written as $[x_1, x_2]$, will be used here to denote the **two-dimensional vector** or **2-vector** OX (see Fig. 9-1).

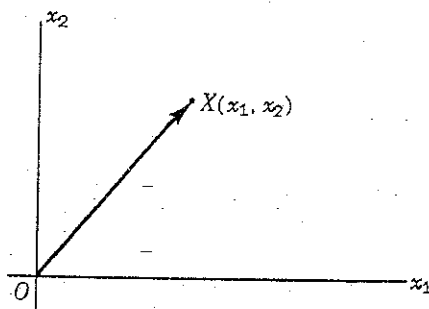


Fig. 9-1

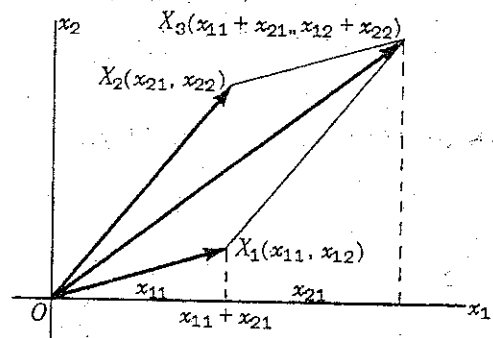


Fig. 9-2

If $X_1 = [x_{11}, x_{12}]$ and $X_2 = [x_{21}, x_{22}]$ are distinct 2-vectors, the parallelogram law for their sum (see Fig. 9-2) yields

$$X_3 = X_1 + X_2 = [x_{11} + x_{21}, x_{12} + x_{22}]$$

Treating X_1 and X_2 as 1×2 matrices, we see that this is merely the rule for adding matrices given in Chapter 1. Moreover, if k is any scalar,

$$kX_1 = [kx_{11}, kx_{12}]$$

is the familiar multiplication of a vector by a real number of physics.

VECTORS. By an n -dimensional vector or n -vector X over F is meant an ordered set of n elements x_i of F , as

$$(9.1) \quad X = [x_1, x_2, \dots, x_n]$$

The elements x_1, x_2, \dots, x_n are called respectively the first, second, ..., n th components of X .

Later we shall find it more convenient to write the components of a vector in a column, as

$$(9.1') \quad X = [x_1, x_2, \dots, x_n]' = \begin{bmatrix} x_1 \\ x_2 \\ \cdot \\ \cdot \\ \cdot \\ x_n \end{bmatrix}$$

Now (9.1) and (9.1') denote the same vector; however, we shall speak of (9.1) as a **row vector** and (9.1') as a **column vector**. We may, then, consider the $p \times q$ matrix A as defining p row vectors (the elements of a row being the components of a q -vector) or as defining q column vectors.

The vector, all of whose components are zero, is called the **zero vector** and is denoted by 0.

The sum and difference of two row (column) vectors and the product of a scalar and a vector are formed by the rules governing matrices.

Example 1. Consider the 3-vectors

$$X_1 = [3, 1, -4], \quad X_2 = [2, 2, -3], \quad X_3 = [0, -4, 1], \quad \text{and} \quad X_4 = [-4, -4, 6]$$

$$(a) \quad 2X_1 - 5X_2 = 2[3, 1, -4] - 5[2, 2, -3] = [6, 2, -8] - [10, 10, -15] = [-4, -8, 7]$$

$$(b) \quad 2X_2 + X_4 = 2[2, 2, -3] + [-4, -4, 6] = [0, 0, 0] = 0$$

$$(c) \quad 2X_1 - 3X_2 - X_3 = 0$$

$$(d) \quad 2X_1 - X_2 - X_3 + X_4 = 0$$

The vectors used here are row vectors. Note that if each bracket is primed to denote column vectors, the results remain correct.

LINEAR DEPENDENCE OF VECTORS. The m n -vectors over F

$$(9.2) \quad \begin{aligned} X_1 &= [x_{11}, x_{12}, \dots, x_{1n}] \\ X_2 &= [x_{21}, x_{22}, \dots, x_{2n}] \\ &\dots\dots\dots \\ X_m &= [x_{m1}, x_{m2}, \dots, x_{mn}] \end{aligned}$$

are said to be **linearly dependent** over F provided there exist m elements k_1, k_2, \dots, k_m of F , not all zero, such that

$$(9.3) \quad k_1X_1 + k_2X_2 + \dots + k_mX_m = 0$$

Otherwise, the m vectors are said to be **linearly independent**.

Example 2. Consider the four vectors of Example 1. By (b) the vectors X_2 and X_4 are linearly dependent; so also are X_1, X_2 , and X_3 by (c) and the entire set by (d).

The vectors X_1 and X_2 , however, are linearly independent. For, assume the contrary so that

$$k_1X_1 + k_2X_2 = [3k_1 + 2k_2, k_1 + 2k_2, -4k_1 - 3k_2] = [0, 0, 0]$$

Then $3k_1 + 2k_2 = 0$, $k_1 + 2k_2 = 0$, and $-4k_1 - 3k_2 = 0$. From the first two relations $k_1 = 0$ and then $k_2 = 0$.

Any n -vector X and the n -zero vector 0 are linearly dependent.

A vector X_{m+1} is said to be expressible as a **linear combination** of the vectors X_1, X_2, \dots, X_m if there exist elements k_1, k_2, \dots, k_m of F such that

$$X_{m+1} = k_1X_1 + k_2X_2 + \dots + k_mX_m$$

BASIC THEOREMS. If in (9.3), $k_i \neq 0$, we may solve for

$$(9.4) \quad \begin{aligned} X_i &= -\frac{1}{k_i} \{k_1X_1 + \dots + k_{i-1}X_{i-1} + k_{i+1}X_{i+1} + \dots + k_mX_m\} && \text{or} \\ X_i &= s_1X_1 + \dots + s_{i-1}X_{i-1} + s_{i+1}X_{i+1} + \dots + s_mX_m \end{aligned}$$

Thus,

I. If m vectors are linearly dependent, some one of them may always be expressed as a linear combination of the others.

II. If m vectors X_1, X_2, \dots, X_m are linearly independent while the set obtained by adding another vector X_{m+1} is linearly dependent, then X_{m+1} can be expressed as a linear combination of X_1, X_2, \dots, X_m .

Example 3. From Example 2, the vectors X_1 and X_2 are linearly independent while X_1, X_2 , and X_3 are linearly dependent, satisfying the relations $2X_1 - 3X_2 - X_3 = 0$. Clearly, $X_3 = 2X_1 - 3X_2$.

III. If among the m vectors X_1, X_2, \dots, X_m there is a subset of $r < m$ vectors which are linearly dependent, the vectors of the entire set are linearly dependent.

Example 4. By (b) of Example 1, the vectors X_2 and X_4 are linearly dependent; by (d), the set of four vectors is linearly dependent. See Problem 1.

IV. If the rank of the matrix

$$(9.5) \quad A = \begin{bmatrix} x_{11} & x_{12} & \cdots & x_{1n} \\ x_{21} & x_{22} & \cdots & x_{2n} \\ \cdots & \cdots & \cdots & \cdots \\ x_{m1} & x_{m2} & \cdots & x_{mn} \end{bmatrix}, \quad m \leq n,$$

associated with the m vectors (9.2) is $r < m$, there are exactly r vectors of the set which are linearly independent while each of the remaining $m-r$ vectors can be expressed as a linear combination of these r vectors. See Problems 2-3.

V. A necessary and sufficient condition that the vectors (9.2) be linearly dependent is that the matrix (9.5) of the vectors be of rank $r < m$. If the rank is m , the vectors are linearly independent.

The set of vectors (9.2) is necessarily linearly dependent if $m > n$.

If the set of vectors (9.2) is linearly independent so also is every subset of them.

A LINEAR FORM over F in n variables x_1, x_2, \dots, x_n is a polynomial of the type

$$(9.6) \quad \sum_{i=1}^n a_i x_i = a_1 x_1 + a_2 x_2 + \cdots + a_n x_n$$

where the coefficients are in F .

Consider a system of m linear forms in n variables

$$(9.7) \quad \begin{cases} f_1 = a_{11}x_1 + a_{12}x_2 + \cdots + a_{1n}x_n \\ f_2 = a_{21}x_1 + a_{22}x_2 + \cdots + a_{2n}x_n \\ \cdots \\ f_m = a_{m1}x_1 + a_{m2}x_2 + \cdots + a_{mn}x_n \end{cases}$$

and the associated matrix

$$A = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \cdots & \cdots & \cdots & \cdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{bmatrix}$$

If there exist elements k_1, k_2, \dots, k_m , not all zero, in F such that

$$k_1 f_1 + k_2 f_2 + \cdots + k_m f_m = 0$$

the forms (9.7) are said to be **linearly dependent**; otherwise the forms are said to be **linearly independent**. Thus, the linear dependence or independence of the forms of (9.7) is equivalent to the linear dependence or independence of the row vectors of A .

Example 5. The forms $f_1 = 2x_1 - x_2 + 3x_3$, $f_2 = x_1 + 2x_2 + 4x_3$, $f_3 = 4x_1 - 7x_2 + x_3$ are linearly dependent since $A = \begin{bmatrix} 2 & -1 & 3 \\ 1 & 2 & 4 \\ 4 & -7 & 1 \end{bmatrix}$ is of rank 2. Here, $3f_1 - 2f_2 - f_3 = 0$.

The system (9.7) is necessarily dependent if $m > n$. Why?

SOLVED PROBLEMS

1. Prove: If among the m vectors X_1, X_2, \dots, X_m there is a subset, say, X_1, X_2, \dots, X_r , $r < m$, which is linearly dependent, so also are the m vectors.

Since, by hypothesis, $k_1X_1 + k_2X_2 + \dots + k_rX_r = 0$ with not all of the k 's equal to zero, then

$$k_1X_1 + k_2X_2 + \dots + k_rX_r + 0 \cdot X_{r+1} + \dots + 0 \cdot X_m = 0$$

with not all of the k 's equal to zero and the entire set of vectors is linearly dependent.

2. Prove: If the rank of the matrix associated with a set of m n -vectors is $r < m$, there are exactly r vectors which are linearly independent while each of the remaining $m-r$ vectors can be expressed as a linear combination of these r vectors.

Let (9.5) be the matrix and suppose first that $m \leq n$. If the r -rowed minor in the upper left hand corner is equal to zero, we interchange rows and columns as are necessary to bring a non-vanishing r -rowed minor into this position and then renumber all rows and columns in natural order. Thus, we have

$$\Delta = \begin{vmatrix} x_{11} & x_{12} & \dots & x_{1r} \\ x_{21} & x_{22} & \dots & x_{2r} \\ \dots & \dots & \dots & \dots \\ x_{r1} & x_{r2} & \dots & x_{rr} \end{vmatrix} \neq 0$$

Consider now an $(r+1)$ -rowed minor

$$\nabla = \begin{vmatrix} x_{11} & x_{12} & \dots & x_{1r} & x_{1q} \\ x_{21} & x_{22} & \dots & x_{2r} & x_{2q} \\ \dots & \dots & \dots & \dots & \dots \\ x_{r1} & x_{r2} & \dots & x_{rr} & x_{rq} \\ x_{p1} & x_{p2} & \dots & x_{pr} & x_{pq} \end{vmatrix} = 0$$

where the elements x_{pj} and x_{iq} are respectively from any row and any column not included in Δ . Let $k_1, k_2, \dots, k_{r+1} = \Delta$ be the respective cofactors of the elements $x_{1q}, x_{2q}, \dots, x_{rq}, x_{pq}$ of the last column of ∇ . Then, by (3.10)

$$k_1x_{1i} + k_2x_{2i} + \dots + k_r x_{ri} + k_{r+1}x_{pi} = 0 \quad (i = 1, 2, \dots, r)$$

and by hypothesis $k_1x_{1q} + k_2x_{2q} + \dots + k_r x_{rq} + k_{r+1}x_{pq} = \nabla = 0$

Now let the last column of ∇ be replaced by another of the remaining columns, say the column numbered u , not appearing in Δ . The cofactors of the elements of this column are precisely the k 's obtained above so that

$$k_1x_{1u} + k_2x_{2u} + \dots + k_r x_{ru} + k_{r+1}x_{pu} = 0$$

Thus,

$$k_1x_{1t} + k_2x_{2t} + \dots + k_r x_{rt} + k_{r+1}x_{pt} = 0 \quad (t = 1, 2, \dots, n)$$

and, summing over all values of t ,

$$k_1X_1 + k_2X_2 + \dots + k_r X_r + k_{r+1}X_p = 0$$

Since $k_{r+1} = \Delta \neq 0$, X_p is a linear combination of the r linearly independent vectors X_1, X_2, \dots, X_r . But X_p was any one of the $m-r$ vectors $X_{r+1}, X_{r+2}, \dots, X_m$; hence, each of these may be expressed as a linear combination of X_1, X_2, \dots, X_r .

For the case $m > n$, consider the matrix when to each of the given m vectors $m-n$ additional zero components are added. This matrix is $[A \mid 0]$. Clearly the linear dependence or independence of the vectors and also the rank of A have not been changed.

Thus, in either case, the vectors X_{r+1}, \dots, X_m are linear combinations of the linearly independent vectors X_1, X_2, \dots, X_r as was to be proved.

3. Show, using a matrix, that each triple of vectors

$$\begin{aligned} X_1 &= [1, 2, -3, 4] & X_1 &= [2, 3, 1, -1] \\ (a) \quad X_2 &= [3, -1, 2, 1] & \text{and} & & (b) \quad X_2 &= [2, 3, 1, -2] \\ X_3 &= [1, -5, 8, -7] & & & X_3 &= [4, 6, 2, -3] \end{aligned}$$

is linearly dependent. In each determine a maximum subset of linearly independent vectors and express the others as linear combinations of these.

(a) Here, $\begin{bmatrix} 1 & 2 & -3 & 4 \\ 3 & -1 & 2 & 1 \\ 1 & -5 & 8 & -7 \end{bmatrix}$ is of rank 2; there are two linearly independent vectors, say X_1 and X_2 . The minor

$\begin{vmatrix} 1 & 2 \\ 3 & -1 \end{vmatrix} \neq 0$. Consider then the minor $\begin{vmatrix} 1 & 2 & -3 \\ 3 & -1 & 2 \\ 1 & -5 & 8 \end{vmatrix}$. The cofactors of the elements of the third column are

respectively $-14, 7,$ and -7 . Then $-14X_1 + 7X_2 - 7X_3 = 0$ and $X_3 = -2X_1 + X_2$.

(b) Here $\begin{bmatrix} 2 & 3 & 1 & -1 \\ 2 & 3 & 1 & -2 \\ 4 & 6 & 2 & -3 \end{bmatrix}$ is of rank 2; there are two linearly independent vectors, say X_1 and X_2 . Now the

minor $\begin{vmatrix} 2 & 3 \\ 2 & 3 \end{vmatrix} = 0$; we interchange the 2nd and 4th columns to obtain $\begin{bmatrix} 2 & -1 & 1 & 3 \\ 2 & -2 & 1 & 3 \\ 4 & -3 & 2 & 6 \end{bmatrix}$ for which $\begin{vmatrix} 2 & -1 \\ 2 & -2 \end{vmatrix} \neq 0$.

The cofactors of the elements of the last column of $\begin{bmatrix} 2 & -1 & 1 \\ 2 & -2 & 1 \\ 4 & -3 & 2 \end{bmatrix}$ are $2, 2, -2$ respectively. Then

$$2X_1 + 2X_2 - 2X_3 = 0 \quad \text{and} \quad X_3 = X_1 + X_2$$

4. Let $P_1(1, 1, 1)$, $P_2(1, 2, 3)$, $P_3(3, 1, 2)$, and $P_4(2, 3, 4)$ be points in ordinary space. The points P_1, P_2 and the origin of coordinates determine a plane π of equation

$$(i) \quad \begin{vmatrix} x & y & z & 1 \\ 1 & 1 & 1 & 1 \\ 1 & 2 & 3 & 1 \\ 0 & 0 & 0 & 1 \end{vmatrix} = x - 2y + z = 0$$

Substituting the coordinates of P_4 into the left member of (i), we have

$$\begin{vmatrix} 2 & 3 & 4 & 1 \\ 1 & 1 & 1 & 1 \\ 1 & 2 & 3 & 1 \\ 0 & 0 & 0 & 1 \end{vmatrix} = \begin{vmatrix} 2 & 3 & 4 & 0 \\ 1 & 1 & 1 & 0 \\ 1 & 2 & 3 & 0 \\ 0 & 0 & 0 & 1 \end{vmatrix} = \begin{vmatrix} 2 & 3 & 4 \\ 1 & 1 & 1 \\ 1 & 2 & 3 \end{vmatrix} = 0$$

Thus, P_4 lies in π . The significant fact here is that $[P_4, P_1, P_2] = \begin{bmatrix} 2 & 3 & 4 \\ 1 & 1 & 1 \\ 1 & 2 & 3 \end{bmatrix}$ is of rank 2.

We have verified: Any three points of ordinary space lie in a plane through the origin provided the matrix of their coordinates is of rank 2.

Show that P_3 does not lie in π .

SUPPLEMENTARY PROBLEMS

5. Prove: If m vectors X_1, X_2, \dots, X_m are linearly independent while the set obtained by adding another vector X_{m+1} is linearly dependent, then X_{m+1} can be expressed as a linear combination of X_1, X_2, \dots, X_m .

6. Show that the representation of X_{m+1} in Problem 5 is unique.

Hint: Suppose $X_{m+1} = \sum_{i=1}^m k_i X_i = \sum_{i=1}^m s_i X_i$ and consider $\sum_{i=1}^m (k_i - s_i) X_i$.

7. Prove: A necessary and sufficient condition that the vectors (9.2) be linearly dependent is that the matrix (9.5) of the vectors be of rank $r < m$.

Hint: Suppose the m vectors are linearly dependent so that (9.4) holds. In (9.5) subtract from the i th row the product of the first row by s_1 , the product of the second row by s_2, \dots as indicated in (9.4). For the converse, see Problem 2.

8. Examine each of the following sets of vectors over the real field for linear dependence or independence. In each dependent set select a maximum linearly independent subset and express each of the remaining vectors as a linear combination of these.

$$(a) \quad \begin{aligned} X_1 &= [2, -1, 3, 2] \\ X_2 &= [1, 3, 4, 2] \\ X_3 &= [3, -5, 2, 2] \end{aligned}$$

$$\text{Ans. (a)} \quad X_3 = 2X_1 - X_2$$

$$(b) \quad \begin{aligned} X_1 &= [1, 2, 1] \\ X_2 &= [2, 1, 4] \\ X_3 &= [4, 5, 6] \\ X_4 &= [1, 8, -3] \end{aligned}$$

$$(b) \quad \begin{aligned} X_3 &= 2X_1 + X_2 \\ X_4 &= 5X_1 - 2X_2 \end{aligned}$$

$$(c) \quad \begin{aligned} X_1 &= [2, 1, 3, 2, -1] \\ X_2 &= [4, 2, 1, -2, 3] \\ X_3 &= [0, 0, 5, 6, -5] \\ X_4 &= [6, 3, -1, -6, 7] \end{aligned}$$

$$(c) \quad \begin{aligned} X_3 &= 2X_1 - X_2 \\ X_4 &= 2X_2 - X_1 \end{aligned}$$

9. Why can there be no more than n linearly independent n -vectors over F ?
10. Show that if in (9.2) either $X_i = X_j$ or $X_i = aX_j$, a in F , the set of vectors is linearly dependent. Is the converse true?
11. Show that any n -vector X and the n -zero vector are linearly dependent; hence, X and 0 are considered proportional. *Hint:* Consider $k_1X + k_2 \cdot 0 = 0$ where $k_1 = 0$ and $k_2 \neq 0$.
12. (a) Show that $X_1 = [1, 1+i, i]$, $X_2 = [i, -i, 1-i]$ and $X_3 = [1+2i, 1-i, 2-i]$ are linearly dependent over the rational field and, hence, over the complex field.
 (b) Show that $X_1 = [1, 1+i, i]$, $X_2 = [i, -i, 1-i]$, and $X_3 = [0, 1-2i, 2-i]$ are linearly independent over the real field but are linearly dependent over the complex field.

13. Investigate the linear dependence or independence of the linear forms:

$$\begin{array}{ll} f_1 = 3x_1 - x_2 + 2x_3 + x_4 & f_1 = 2x_1 - 3x_2 + 4x_3 - 2x_4 \\ (a) f_2 = 2x_1 + 3x_2 - x_3 + 2x_4 & (b) f_2 = 3x_1 + 2x_2 - 2x_3 + 5x_4 \\ f_3 = 5x_1 - 9x_2 + 8x_3 - x_4 & f_3 = 5x_1 - x_2 + 2x_3 + x_4 \end{array}$$

Ans. (a) $3f_1 - 2f_2 - f_3 = 0$

14. Consider the linear dependence or independence of a system of polynomials

$$P_i = a_{i0}x^n + a_{i1}x^{n-1} + \dots + a_{in-1}x + a_{in} \quad (i = 1, 2, \dots, m)$$

and show that the system is linearly dependent or independent according as the row vectors of the coefficient matrix

$$A = \begin{bmatrix} a_{10} & a_{11} & \dots & a_{1n} \\ a_{20} & a_{21} & \dots & a_{2n} \\ \dots & \dots & \dots & \dots \\ a_{m0} & a_{m1} & \dots & a_{mn} \end{bmatrix}$$

are linearly dependent or independent, that is, according as the rank r of A is less than or equal to m .

15. If the polynomials of either system are linearly dependent, find a linear combination which is identically zero.

$$\begin{array}{ll} P_1 = x^3 - 3x^2 + 4x - 2 & P_1 = 2x^4 + 3x^3 - 4x^2 + 5x + 3 \\ (a) P_2 = 2x^2 - 6x + 4 & (b) P_2 = x^3 + 2x^2 - 3x + 1 \\ P_3 = x^3 - 2x^2 + x & P_3 = x^4 + 2x^3 - x^2 + x + 2 \end{array}$$

Ans. (a) $2P_1 + P_2 - 2P_3 = 0$ (b) $P_1 + P_2 - 2P_3 = 0$

16. Consider the linear dependence or independence of a set of 2×2 matrices $M_1 = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$, $M_2 = \begin{bmatrix} e & f \\ g & h \end{bmatrix}$, $M_3 = \begin{bmatrix} p & q \\ s & t \end{bmatrix}$ over F .

Show that $k_1M_1 + k_2M_2 + k_3M_3 = 0$, when not all the k 's (in F) are zero, requires that the rank of the

matrix $\begin{bmatrix} a & b & c & d \\ e & f & g & h \\ p & q & s & t \end{bmatrix}$ be < 3 . (Note that the matrices M_1, M_2, M_3 are considered as defining vectors of four components.)

Extend the result to a set of $m \times n$ matrices.

17. Show that $\begin{bmatrix} 1 & 2 & 3 \\ 3 & 2 & 4 \\ 1 & 3 & 2 \end{bmatrix}$, $\begin{bmatrix} 2 & 1 & 3 \\ 3 & 4 & 2 \\ 2 & 2 & 1 \end{bmatrix}$, and $\begin{bmatrix} 0 & 3 & 3 \\ 3 & 0 & 6 \\ 0 & 4 & 3 \end{bmatrix}$ are linearly dependent.

18. Show that any 2×2 matrix can be written as a linear combination of the matrices $\begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$, $\begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$, $\begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}$, and $\begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}$. Generalize to $n \times n$ matrices.

19. If the n -vectors X_1, X_2, \dots, X_n are linearly independent, show that the vectors Y_1, Y_2, \dots, Y_n , where $Y_i = \sum_{j=1}^n a_{ij} X_j$, are linearly independent if and only if $A = [a_{ij}]$ is non-singular.

20. If A is of rank r , show how to construct a non-singular matrix B such that $AB = [C_1, C_2, \dots, C_r, 0, \dots, 0]$ where C_1, C_2, \dots, C_r are a given set of linearly independent columns of A .

21. Given the points $P_1(1, 1, 1, 1)$, $P_2(1, 2, 3, 4)$, $P_3(2, 2, 2, 2)$, and $P_4(3, 4, 5, 6)$ of four-dimensional space.
 (a) Show that the rank of $[P_1, P_3]'$ is 1 so that the points lie on a line through the origin.
 (b) Show that $[P_1, P_2, P_3, P_4]'$ is of rank 2 so that these points lie in a plane through the origin.
 (c) Does $P_5(2, 3, 2, 5)$ lie in the plane of (b)?

22. Show that every n -square matrix A over F satisfies an equation of the form

$$A^p + k_1 A^{p-1} + k_2 A^{p-2} + \dots + k_{p-1} A + k_p I = 0$$

where the k_i are scalars of F .

Hint: Consider $I, A, A^2, A^3, \dots, A^{n^2}$ in the light of Problem 16.

23. Find the equation of minimum degree (see Problem 22) which is satisfied by

$$(a) A = \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}, \quad (b) A = \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix}, \quad (c) A = \begin{bmatrix} 1 & 0 \\ 1 & 1 \end{bmatrix}$$

Ans. (a) $A^2 - 2A = 0$, (b) $A^2 - 2A + 2I = 0$, (c) $A^2 - 2A + I = 0$

24. In Problem 23(b) and (c), multiply each equation by A^{-1} to obtain (b) $A^{-1} = I - \frac{1}{2}A$, (c) $A^{-1} = 2I - A$, and thus verify: If A over F is non-singular, then A^{-1} can be expressed as a polynomial in A whose coefficients are scalars of F .

To solve the system (10.1) by means of (10.4), we proceed by elementary row transformations to replace A by the row equivalent canonical matrix of Chapter 5. In doing this, we operate on the entire rows of (10.4).

Example 1. Solve the system

$$\begin{cases} x_1 + 2x_2 + x_3 = 2 \\ 3x_1 + x_2 - 2x_3 = 1 \\ 4x_1 - 3x_2 - x_3 = 3 \\ 2x_1 + 4x_2 + 2x_3 = 4 \end{cases}$$

The augmented matrix $[A \ H] = \begin{bmatrix} 1 & 2 & 1 & 2 \\ 3 & 1 & -2 & 1 \\ 4 & -3 & -1 & 3 \\ 2 & 4 & 2 & 4 \end{bmatrix} \sim \begin{bmatrix} 1 & 2 & 1 & 2 \\ 0 & -5 & -5 & -5 \\ 0 & -11 & -5 & -5 \\ 0 & 0 & 0 & 0 \end{bmatrix} \sim \begin{bmatrix} 1 & 2 & 1 & 2 \\ 0 & 1 & 1 & 1 \\ 0 & -11 & -5 & -5 \\ 0 & 0 & 0 & 0 \end{bmatrix}$

$$\sim \begin{bmatrix} 1 & 0 & -1 & 0 \\ 0 & 1 & 1 & 1 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix} \sim \begin{bmatrix} 1 & 0 & 0 & 1 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

Thus, the solution is the equivalent system of equations: $x_1 = 1$, $x_2 = 0$, $x_3 = 1$. Expressed in vector form, we have $X = [1, 0, 1]^t$.

FUNDAMENTAL THEOREMS. When the coefficient matrix A of the system (10.1) is reduced to the row equivalent canonical form C , suppose $[A \ H]$ is reduced to $[C \ K]$, where $K = [k_1, k_2, \dots, k_m]^t$. If A is of rank r , the first r rows of C contain one or more non-zero elements. The first non-zero element in each of these rows is 1 and the column in which that 1 stands has zeros elsewhere. The remaining rows consist of zeros. From the first r rows of $[C \ K]$, we may obtain each of the variables $x_{j_1}, x_{j_2}, \dots, x_{j_r}$ (the notation is that of Chapter 5) in terms of the remaining variables $x_{j_{r+1}}, x_{j_{r+2}}, \dots, x_{j_n}$ and one of the k_1, k_2, \dots, k_r .

If $k_{r+1} = k_{r+2} = \dots = k_m = 0$, then (10.1) is consistent and an arbitrarily selected set of values for $x_{j_{r+1}}, x_{j_{r+2}}, \dots, x_{j_n}$ together with the resulting values of $x_{j_1}, x_{j_2}, \dots, x_{j_r}$ constitute a solution. On the other hand, if at least one of $k_{r+1}, k_{r+2}, \dots, k_m$ is different from zero, say $k_t \neq 0$, the corresponding equation reads

$$0x_1 + 0x_2 + \dots + 0x_n = k_t \neq 0$$

and (10.1) is inconsistent.

In the consistent case, A and $[A \ H]$ have the same rank; in the inconsistent case, they have different ranks. Thus

I. A system $AX = H$ of m linear equations in n unknowns is consistent if and only if the coefficient matrix and the augmented matrix of the system have the same rank.

II. In a consistent system (10.1) of rank $r < n$, $n - r$ of the unknowns may be chosen so that the coefficient matrix of the remaining r unknowns is of rank r . When these $n - r$ unknowns are assigned any values whatever, the other r unknowns are uniquely determined.

Example 2. For the system

$$\begin{cases} x_1 + 2x_2 - 3x_3 - 4x_4 = 6 \\ x_1 + 3x_2 + x_3 - 2x_4 = 4 \\ 2x_1 + 5x_2 - 2x_3 - 5x_4 = 10 \end{cases}$$

$$\begin{aligned}
 [A \ H] &= \left[\begin{array}{cccc|c} 1 & 2 & -3 & -4 & 6 \\ 1 & 3 & 1 & -2 & 4 \\ 2 & 5 & -2 & -5 & 10 \end{array} \right] \sim \left[\begin{array}{cccc|c} 1 & 2 & -3 & -4 & 6 \\ 0 & 1 & 4 & 2 & -2 \\ 0 & 1 & 4 & 3 & -2 \end{array} \right] \sim \left[\begin{array}{cccc|c} 1 & 0 & -11 & -8 & 10 \\ 0 & 1 & 4 & 2 & -2 \\ 0 & 0 & 0 & 1 & 0 \end{array} \right] \\
 &\sim \left[\begin{array}{cccc|c} 1 & 0 & -11 & 0 & 10 \\ 0 & 1 & 4 & 0 & -2 \\ 0 & 0 & 0 & 1 & 0 \end{array} \right] = [C \ K]
 \end{aligned}$$

Since A and $[A \ H]$ are each of rank $r = 3$, the given system is consistent; moreover, the general solution contains $n - r = 4 - 3 = 1$ arbitrary constants. From the last row of $[C \ K]$, $x_4 = 0$. Let $x_3 = a$, where a is arbitrary; then $x_1 = 10 + 11a$ and $x_2 = -2 - 4a$. The solution of the system is given by $x_1 = 10 + 11a$, $x_2 = -2 - 4a$, $x_3 = a$, $x_4 = 0$ or $X = [10 + 11a, -2 - 4a, a, 0]^T$.

If a consistent system of equations over F has a unique solution (Example 1) that solution is over F . If the system has infinitely many solutions (Example 2) it has infinitely many solutions over F when the arbitrary values to be assigned are over F . However, the system has infinitely many solutions over any field \mathfrak{F} of which F is a subfield. For example, the system of Example 2 has infinitely many solutions over F (the rational field) if a is restricted to rational numbers, it has infinitely many real solutions if a 's restricted to real numbers, it has infinitely many complex solutions if a is any complex number whatever.

See Problems 1-2.

NON-HOMOGENEOUS EQUATIONS. A linear equation

$$a_1x_1 + a_2x_2 + \dots + a_nx_n = h$$

is called **non-homogeneous** if $h \neq 0$. A system $AX = H$ is called a system of non-homogeneous equations provided H is not a zero vector. The systems of Examples 1 and 2 are non-homogeneous systems.

In Problem 3 we prove

III. A system of n non-homogeneous equations in n unknowns has a unique solution provided the rank of its coefficient matrix A is n , that is, provided $|A| \neq 0$.

In addition to the method above, two additional procedures for solving a consistent system of n non-homogeneous equations in as many unknowns $AX = H$ are given below. The first of these is the familiar solution by determinants.

(a) Solution by Cramer's Rule. Denote by A_i , ($i = 1, 2, \dots, n$) the matrix obtained from A by replacing its i th column with the column of constants (the h 's). Then, if $|A| \neq 0$, the system $AX = H$ has the unique solution

$$(10.5) \quad x_1 = \frac{|A_1|}{|A|}, \quad x_2 = \frac{|A_2|}{|A|}, \quad \dots, \quad x_n = \frac{|A_n|}{|A|}$$

See Problem 4.

Example 3. Solve the system

$$\begin{cases} 2x_1 + x_2 + 5x_3 + x_4 = 5 \\ x_1 + x_2 - 3x_3 - 4x_4 = -1 \\ 3x_1 + 6x_2 - 2x_3 + x_4 = 8 \\ 2x_1 + 2x_2 + 2x_3 - 3x_4 = 2 \end{cases}$$

using Cramer's Rule.

We find

$$|A| = \begin{vmatrix} 2 & 1 & 5 & 1 \\ 1 & 1 & -3 & -4 \\ 3 & 6 & -2 & 1 \\ 2 & 2 & 2 & -3 \end{vmatrix} = -120, \quad |A_1| = \begin{vmatrix} 5 & 1 & 5 & 1 \\ -1 & 1 & -3 & -4 \\ 8 & 6 & -2 & 1 \\ 2 & 2 & 2 & -3 \end{vmatrix} = -240$$

$$|A_2| = \begin{vmatrix} 2 & 5 & 5 & 1 \\ 1 & -1 & -3 & -4 \\ 3 & 8 & -2 & 1 \\ 2 & 2 & 2 & -3 \end{vmatrix} = -24, \quad |A_3| = \begin{vmatrix} 2 & 1 & 5 & 1 \\ 1 & 1 & -1 & -4 \\ 3 & 6 & 8 & 1 \\ 2 & 2 & 2 & -3 \end{vmatrix} = 0$$

and

$$|A_4| = \begin{vmatrix} 2 & 1 & 5 & 5 \\ 1 & 1 & -3 & -1 \\ 3 & 6 & -2 & 8 \\ 2 & 2 & 2 & 2 \end{vmatrix} = -96$$

Then $x_1 = \frac{|A_1|}{|A|} = \frac{-240}{-120} = 2$, $x_2 = \frac{|A_2|}{|A|} = \frac{-24}{-120} = \frac{1}{5}$, $x_3 = \frac{|A_3|}{|A|} = \frac{0}{-120} = 0$, and

$$x_4 = \frac{|A_4|}{|A|} = \frac{-96}{-120} = \frac{4}{5}.$$

(b) Solution using A^{-1} . If $|A| \neq 0$, A^{-1} exists and the solution of the system $AX = H$ is given by

$$(10.6) \quad A^{-1} \cdot AX = A^{-1}H \quad \text{or} \quad X = A^{-1}H$$

Example 4. The coefficient matrix of the system
$$\begin{cases} 2x_1 + 3x_2 + x_3 = 9 \\ x_1 + 2x_2 + 3x_3 = 6 \\ 3x_1 + x_2 + 2x_3 = 8 \end{cases}$$
 is $A = \begin{bmatrix} 2 & 3 & 1 \\ 1 & 2 & 3 \\ 3 & 1 & 2 \end{bmatrix}$.

From Problem 2(b), Chapter 7, $A^{-1} = \frac{1}{18} \begin{bmatrix} 1 & -5 & 7 \\ 7 & 1 & -5 \\ -5 & 7 & 1 \end{bmatrix}$. Then

$$A^{-1} \cdot AX = X = A^{-1}H = \frac{1}{18} \begin{bmatrix} 1 & -5 & 7 \\ 7 & 1 & -5 \\ -5 & 7 & 1 \end{bmatrix} \begin{bmatrix} 9 \\ 6 \\ 8 \end{bmatrix} = \frac{1}{18} \begin{bmatrix} 35 \\ 29 \\ 5 \end{bmatrix}$$

The solution of the system is $x_1 = 35/18$, $x_2 = 29/18$, $x_3 = 5/18$.

See Problem 5.

HOMOGENEOUS EQUATIONS. A linear equation

$$(10.7) \quad a_1x_1 + a_2x_2 + \dots + a_nx_n = 0$$

is called **homogeneous**. A system of linear equations

$$(10.8) \quad AX = 0$$

in n unknowns is called a **system of homogeneous equations**. For the system (10.8) the rank of the coefficient matrix A and the augmented matrix $[A \ 0]$ are the same; thus, the system is always consistent. Note that $X = 0$, that is, $x_1 = x_2 = \dots = x_n = 0$ is always a solution; it is called the **trivial solution**.

If the rank of A is n , then n of the equations of (10.8) can be solved by Cramer's rule for the unique solution $x_1 = x_2 = \dots = x_n = 0$ and the system has only the trivial solution. If the rank of A is $r < n$, Theorem II assures the existence of non-trivial solutions. Thus,

IV. A necessary and sufficient condition for (10.8) to have a solution other than the trivial solution is that the rank of A be $r < n$.

V. A necessary and sufficient condition that a system of n homogeneous equations in n unknowns has a solution other than the trivial solution is $|A| = 0$.

VI. If the rank of (10.8) is $r < n$, the system has exactly $n-r$ linearly independent solutions such that every solution is a linear combination of these $n-r$ and every such linear combination is a solution.

See Problem 6.

LET X_1 and X_2 be two distinct solutions of $AX = H$. Then $AX_1 = H$, $AX_2 = H$, and $A(X_1 - X_2) = AY = 0$. Thus, $Y = X_1 - X_2$ is a non-trivial solution of $AX = 0$.

Conversely, if Z is any non-trivial solution of $AX = 0$ and if X_p is any solution of $AX = H$, then $X = X_p + Z$ is also a solution of $AX = H$. As Z ranges over the complete solution of $AX = 0$, $X_p + Z$ ranges over the complete solution of $AX = H$. Thus,

VII. If the system of non-homogeneous equations $AX = H$ is consistent, a complete solution of the system is given by the complete solution of $AX = 0$ plus any particular solution of $AX = H$.

Example 5. In the system $\begin{cases} x_1 - 2x_2 + 3x_3 = 4 \\ x_1 + x_2 + 2x_3 = 5 \end{cases}$ set $x_1 = 0$; then $x_3 = 2$ and $x_2 = 1$. A particular

solution is $X_p = [0, 1, 2]^T$. The complete solution of $\begin{cases} x_1 - 2x_2 + 3x_3 = 0 \\ x_1 + x_2 + 2x_3 = 0 \end{cases}$ is $[-7a, a, 3a]^T$,

where a is arbitrary. Then the complete solution of the given system is

$$X = [-7a, a, 3a]^T + [0, 1, 2]^T = [-7a, 1+a, 2+3a]^T$$

Note. The above procedure may be extended to larger systems. However, it is first necessary to show that the system is consistent. This is a long step in solving the system by the augmented matrix method given earlier.

SOLVED PROBLEMS

1. Solve $\begin{cases} x_1 + x_2 - 2x_3 + x_4 + 3x_5 = 1 \\ 2x_1 - x_2 + 2x_3 + 2x_4 + 6x_5 = 2 \\ 3x_1 + 2x_2 - 4x_3 - 3x_4 - 9x_5 = 3 \end{cases}$

Solution:

The augmented matrix

$$[A \ H] = \left[\begin{array}{ccccc|c} 1 & 1 & -2 & 1 & 3 & 1 \\ 2 & -1 & 2 & 2 & 6 & 2 \\ 3 & 2 & -4 & -3 & -9 & 3 \end{array} \right] \sim \left[\begin{array}{ccccc|c} 1 & 1 & -2 & 1 & 3 & 1 \\ 0 & -3 & 6 & 0 & 0 & 0 \\ 0 & -1 & 2 & -6 & -18 & 0 \end{array} \right] \sim \left[\begin{array}{ccccc|c} 1 & 1 & -2 & 1 & 3 & 1 \\ 0 & 1 & -2 & 0 & 0 & 0 \\ 0 & -1 & 2 & -6 & -18 & 0 \end{array} \right]$$

$$\sim \left[\begin{array}{ccccc|c} 1 & 0 & 0 & 1 & 3 & 1 \\ 0 & 1 & -2 & 0 & 0 & 0 \\ 0 & 0 & 0 & -6 & -18 & 0 \end{array} \right] \sim \left[\begin{array}{ccccc|c} 1 & 0 & 0 & 1 & 3 & 1 \\ 0 & 1 & -2 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 3 & 0 \end{array} \right]$$

$$\sim \left[\begin{array}{ccccc|c} 1 & 0 & 0 & 0 & 0 & 1 \\ 0 & 1 & -2 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 3 & 0 \end{array} \right]$$

Then $x_1 = 1$, $x_2 - 2x_3 = 0$, and $x_4 + 3x_5 = 0$. Take $x_3 = a$ and $x_5 = b$, where a and b are arbitrary; the complete solution may be given as $x_1 = 1$, $x_2 = 2a$, $x_3 = a$, $x_4 = -3b$, $x_5 = b$ or as $X = [1, 2a, a, -3b, b]^T$.

2. Solve $\begin{cases} x_1 + x_2 + 2x_3 + x_4 = 5 \\ 2x_1 + 3x_2 - x_3 - 2x_4 = 2 \\ 4x_1 + 5x_2 + 3x_3 = 7 \end{cases}$

Solution:

$$[A \ H] = \left[\begin{array}{cccc|c} 1 & 1 & 2 & 1 & 5 \\ 2 & 3 & -1 & -2 & 2 \\ 4 & 5 & 3 & 0 & 7 \end{array} \right] \sim \left[\begin{array}{cccc|c} 1 & 1 & 2 & 1 & 5 \\ 0 & 1 & -5 & -4 & -8 \\ 0 & 1 & -5 & -4 & -13 \end{array} \right] \sim \left[\begin{array}{cccc|c} 1 & 0 & 7 & 5 & 13 \\ 0 & 1 & -5 & -4 & -8 \\ 0 & 0 & 0 & 0 & -5 \end{array} \right]$$

The last row reads $0 \cdot x_1 + 0 \cdot x_2 + 0 \cdot x_3 + 0 \cdot x_4 = -5$; thus the given system is inconsistent and has no solution.

$$X = \frac{1}{120} \begin{bmatrix} 120 & 120 & 0 & -120 \\ -69 & -73 & 17 & 80 \\ -15 & -35 & -5 & 40 \\ 24 & 8 & 8 & -40 \end{bmatrix} \begin{bmatrix} 5 \\ -1 \\ 8 \\ 2 \end{bmatrix} = \begin{bmatrix} 2 \\ 1/5 \\ 0 \\ 4/5 \end{bmatrix}$$

(See Example 3.)

6. Solve
$$\begin{cases} x_1 + x_2 + x_3 + x_4 = 0 \\ x_1 + 3x_2 + 2x_3 + 4x_4 = 0 \\ 2x_1 + x_3 - x_4 = 0 \end{cases}$$

Solution:

$$\begin{aligned} [A \ H] &= \begin{bmatrix} 1 & 1 & 1 & 1 & 0 \\ 1 & 3 & 2 & 4 & 0 \\ 2 & 0 & 1 & -1 & 0 \end{bmatrix} \sim \begin{bmatrix} 1 & 1 & 1 & 1 & 0 \\ 0 & 2 & 1 & 3 & 0 \\ 0 & -2 & -1 & -3 & 0 \end{bmatrix} \sim \begin{bmatrix} 1 & 1 & 1 & 1 & 0 \\ 0 & 2 & 1 & 3 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \\ &\sim \begin{bmatrix} 1 & 1 & 1 & 1 & 0 \\ 0 & 1 & \frac{1}{2} & \frac{3}{2} & 0 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \sim \begin{bmatrix} 1 & 0 & \frac{1}{2} & -\frac{1}{2} & 0 \\ 0 & 1 & \frac{1}{2} & \frac{3}{2} & 0 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \end{aligned}$$

The complete solution of the system is $x_1 = -\frac{1}{2}a + \frac{1}{2}b$, $x_2 = -\frac{1}{2}a - \frac{3}{2}b$, $x_3 = a$, $x_4 = b$. Since the rank of A is 2, we may obtain exactly $n-r = 4-2 = 2$ linearly independent solutions. One such pair, obtained by first taking $a = 1$, $b = 1$ and then $a = 3$, $b = 1$ is

$$x_1 = 0, x_2 = -2, x_3 = 1, x_4 = 1 \quad \text{and} \quad x_1 = -1, x_2 = -3, x_3 = 3, x_4 = 1$$

What can be said of the pair of solutions obtained by taking $a = b = 1$ and $a = b = 3$?

7. Prove: In a square matrix A of order n and rank $n-1$, the cofactors of the elements of any two rows (columns) are proportional.

Since $|A| = 0$, the cofactors of the elements of any row (column) of A are a solution X_1 of the system $AX = 0$ ($A^T X = 0$).

Now the system has but one linearly independent solution since A (A^T) is of rank $n-1$. Hence, for the cofactors of another row (column) of A (another solution X_2 of the system), we have $X_2 = kX_1$.

8. Prove: If f_1, f_2, \dots, f_m are $m < n$ linearly independent linear forms over F in n variables, then the p linear forms

$$g_j = \sum_{i=1}^m s_{ij} f_i, \quad (j = 1, 2, \dots, p)$$

are linearly dependent if and only if the $m \times p$ matrix $[s_{ij}]$ is of rank $r < p$.

The g 's are linearly dependent if and only if there exist scalars a_1, a_2, \dots, a_p in F , not all zero, such that

$$\begin{aligned} a_1 g_1 + a_2 g_2 + \dots + a_p g_p &= a_1 \sum_{i=1}^m s_{i1} f_i + a_2 \sum_{i=1}^m s_{i2} f_i + \dots + a_p \sum_{i=1}^m s_{ip} f_i \\ &= \left(\sum_{j=1}^p a_j s_{1j} \right) f_1 + \left(\sum_{j=1}^p a_j s_{2j} \right) f_2 + \dots + \left(\sum_{j=1}^p a_j s_{mj} \right) f_m \\ &= \sum_{i=1}^m \left(\sum_{j=1}^p a_j s_{ij} \right) f_i = 0 \end{aligned}$$

12. Reconcile the solution of 10(d) with another $x_1 = c, x_2 = d, x_3 = -\frac{10}{3}c - \frac{d}{3}, x_4 = \frac{8}{3}c + \frac{5}{3}d$.

13. Given $A = \begin{bmatrix} 1 & 1 & 2 \\ 2 & 2 & 4 \\ 3 & 3 & 6 \end{bmatrix}$, find a matrix B of rank 2 such that $AB = 0$. Hint. Select the columns of B from the solutions of $AX = 0$.

14. Show that a square matrix is singular if and only if its rows (columns) are linearly dependent.

15. Let $AX = 0$ be a system of n homogeneous equations in n unknowns and suppose A of rank $r = n - 1$. Show that any non-zero vector of cofactors $[\alpha_{i1}, \alpha_{i2}, \dots, \alpha_{in}]^t$ of a row of A is a solution of $AX = 0$.

16. Use Problem 15 to solve:

$$(a) \begin{cases} x_1 - 2x_2 + 3x_3 = 0 \\ 2x_1 + 5x_2 + 6x_3 = 0 \end{cases} \quad (b) \begin{cases} 2x_1 + 3x_2 - x_3 = 0 \\ 3x_1 - 4x_2 + 2x_3 = 0 \end{cases} \quad (c) \begin{cases} 2x_1 + 3x_2 + 4x_3 = 0 \\ 2x_1 - x_2 + 6x_3 = 0 \end{cases}$$

Hint. To the equations of (a) adjoin $0x_1 + 0x_2 + 0x_3 = 0$ and find the cofactors of the elements of the

third row of $\begin{bmatrix} 1 & -2 & 3 \\ 2 & 5 & 6 \\ 0 & 0 & 0 \end{bmatrix}$.

Ans: (a) $x_1 = -27a, x_2 = 0, x_3 = 9a$ or $[3a, 0, -a]^t$, (b) $[2a, -7a, -17a]^t$, (c) $[11a, -2a, -4a]^t$

17. Let the coefficient and the augmented matrix of the system of 3 non-homogeneous equations in 5 unknowns $AX = H$ be of rank 2 and assume the canonical form of the augmented matrix to be

$$\begin{bmatrix} 1 & 0 & b_{13} & b_{14} & b_{15} & c_1 \\ 0 & 1 & b_{23} & b_{24} & b_{25} & c_2 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

with not both of c_1, c_2 equal to 0. First choose $x_3 = x_4 = x_5 = 0$ and obtain $X_1 = [c_1, c_2, 0, 0, 0]^t$ as a solution of $AX = H$. Then choose $x_3 = 1, x_4 = x_5 = 0$, also $x_3 = x_5 = 0, x_4 = 1$ and $x_3 = x_4 = 0, x_5 = 1$ to obtain other solutions X_2, X_3 , and X_4 . Show that these $5 - 2 + 1 = 4$ solutions are linearly independent.

18. Consider the linear combination $Y = s_1X_1 + s_2X_2 + s_3X_3 + s_4X_4$ of the solutions of Problem 17. Show that Y is a solution of $AX = H$ if and only if (i) $s_1 + s_2 + s_3 + s_4 = 1$. Thus, with s_1, s_2, s_3, s_4 arbitrary except for (i), Y is a complete solution of $AX = H$.

19. Prove: Theorem VI. Hint. Follow Problem 17 with $c_1 = c_2 = 0$.

20. Prove: If A is an $m \times p$ matrix of rank r_1 and B is a $p \times n$ matrix of rank r_2 such that $AB = 0$, then $r_1 + r_2 \leq p$. Hint. Use Theorem VI.

21. Using the 4×5 matrix $A = [a_{ij}]$ of rank 2, verify: In an $m \times n$ matrix A of rank r , the r -square determinants formed from the columns of a submatrix consisting of any r rows of A are proportional to the r -square determinants formed from any other submatrix consisting of r rows of A .

Hint. Suppose the first two rows are linearly independent so that $a_{3j} = p_{31}a_{1j} + p_{32}a_{2j}, a_{4j} = p_{41}a_{1j} + p_{42}a_{2j}, (j = 1, 2, \dots, 5)$. Evaluate the 2-square determinants

$$\begin{vmatrix} a_{1q} & a_{1s} \\ a_{2q} & a_{2s} \end{vmatrix}, \quad \begin{vmatrix} a_{1q} & a_{1s} \\ a_{3q} & a_{3s} \end{vmatrix}, \quad \text{and} \quad \begin{vmatrix} a_{3q} & a_{3s} \\ a_{4q} & a_{4s} \end{vmatrix}$$

22. Write a proof of the theorem of Problem 21.

23. From Problem 7, obtain: If the n -square matrix A is of rank $n - 1$, then the following relations among its cofactors hold

$$(a) \alpha_{ij} \alpha_{hk} = \alpha_{ik} \alpha_{hj}, \quad (b) \alpha_{ii} \alpha_{jj} = \alpha_{ij} \alpha_{ji}$$

where $(h, i, j, k = 1, 2, \dots, n)$.

24. Show that $B = \begin{bmatrix} 1 & 1 & 1 & 1 & 4 \\ 1 & 2 & 3 & -4 & 2 \\ 2 & 1 & 1 & 2 & 6 \\ 3 & 2 & -1 & -1 & 3 \\ 1 & 2 & 2 & -2 & 4 \\ 2 & 3 & -3 & 1 & 1 \end{bmatrix}$ is row equivalent to $\begin{bmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}$. From $B = [A \ H]$ infer that the

system of 6 linear equations in 4 unknowns has 5 linearly independent equations. Show that a system of $m > n$ linear equations in n unknowns can have at most $n + 1$ linearly independent equations. Show that when there are $n + 1$, the system is inconsistent.

25. If $AX = H$ is consistent and of rank r , for what set of r variables can one solve?

26. Generalize the results of Problems 17 and 18 to m non-homogeneous equations in n unknowns with coefficient and augmented matrix of the same rank r to prove: If the coefficient and the augmented matrix of the system $AX = H$ of m non-homogeneous equations in n unknowns have rank r and if $X_1, X_2, \dots, X_{n-r+1}$ are linearly independent solutions of the system, then

$$X = s_1 X_1 + s_2 X_2 + \dots + s_{n-r+1} X_{n-r+1}$$

where $\sum_{i=1}^{n-r+1} s_i = 1$, is a complete solution.

27. In a four-pole electrical network, the input quantities E_1 and I_1 are given in terms of the output quantities E_2 and I_2 by

$$\begin{aligned} E_1 &= aE_2 + bI_2 \\ I_1 &= cE_2 + dI_2 \end{aligned} \quad \text{or} \quad \begin{bmatrix} E_1 \\ I_1 \end{bmatrix} = \begin{bmatrix} a & b \\ c & d \end{bmatrix} \begin{bmatrix} E_2 \\ I_2 \end{bmatrix} = A \begin{bmatrix} E_2 \\ I_2 \end{bmatrix}$$

Show that $\begin{bmatrix} E_1 \\ E_2 \end{bmatrix} = \frac{1}{c} \begin{bmatrix} a & -|A| \\ 1 & -d \end{bmatrix} \begin{bmatrix} I_1 \\ I_2 \end{bmatrix}$ and $\begin{bmatrix} E_1 \\ I_2 \end{bmatrix} = \frac{1}{d} \begin{bmatrix} b & |A| \\ 1 & -c \end{bmatrix} \begin{bmatrix} I_1 \\ E_2 \end{bmatrix}$.

Solve also for E_2 and I_2 , I_1 and I_2 , I_1 and E_2 .

28. Let the system of n linear equations in n unknowns $AX = H$, $H \neq 0$, have a unique solution. Show that the system $AX = K$ has a unique solution for any n -vector $K \neq 0$.

29. Solve the set of linear forms $AX = \begin{bmatrix} 1 & -1 & 1 \\ 2 & 1 & 3 \\ 1 & 2 & 3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = Y = \begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix}$ for the x_i as linear forms in the y 's.

Now write down the solution of $A'X = Y$.

30. Let A be n -square and non-singular, and let S_i be the solution of $AX = E_i$, ($i = 1, 2, \dots, n$), where E_i is the n -vector whose i th component is 1 and whose other components are 0. Identify the matrix $[S_1, S_2, \dots, S_n]$.

31. Let A be an $m \times n$ matrix with $m < n$ and let S_i be a solution of $AX = E_i$, ($i = 1, 2, \dots, m$), where E_i is the m -vector whose i th component is 1 and whose other components are 0. If $K = [k_1, k_2, \dots, k_m]'$, show that

$$k_1 S_1 + k_2 S_2 + \dots + k_m S_m$$

is a solution of $AX = K$.

Chapter 11

Vector Spaces

UNLESS STATED OTHERWISE, all vectors will now be column vectors. When components are displayed, we shall write $[x_1, x_2, \dots, x_n]'$. The transpose mark ($'$) indicates that the elements are to be written in a column.

A set of such n -vectors over F is said to be **closed under addition** if the sum of any two of them is a vector of the set. Similarly, the set is said to be **closed under scalar multiplication** if every scalar multiple of a vector of the set is a vector of the set.

Example 1. (a) The set of all vectors $[x_1, x_2, x_3]'$ of ordinary space having equal components ($x_1 = x_2 = x_3$) is closed under both addition and scalar multiplication. For, the sum of any two of the vectors and k times any vector (k real) are again vectors having equal components.

(b) The set of all vectors $[x_1, x_2, x_3]'$ of ordinary space is closed under addition and scalar multiplication.

VECTOR SPACES. Any set of n -vectors over F which is closed under both addition and scalar multiplication is called a **vector space**. Thus, if X_1, X_2, \dots, X_m are n -vectors over F , the set of all linear combinations

$$(11.1) \quad k_1 X_1 + k_2 X_2 + \dots + k_m X_m \quad (k_i \text{ in } F)$$

is a vector space over F . For example, both of the sets of vectors (a) and (b) of Example 1 are vector spaces. Clearly, every vector space (11.1) contains the zero n -vector while the zero n -vector alone is a vector space. (The space (11.1) is also called a **linear vector space**.)

The totality $V_n(F)$ of all n -vectors over F is called the **n -dimensional vector space over F** .

SUBSPACES. A set V of the vectors of $V_n(F)$ is called a **subspace** of $V_n(F)$ provided V is closed under addition and scalar multiplication. Thus, the zero n -vector is a subspace of $V_n(F)$; so also is $V_n(F)$ itself. The set (a) of Example 1 is a subspace (a line) of ordinary space. In general, if X_1, X_2, \dots, X_m belong to $V_n(F)$, the space of all linear combinations (11.1) is a subspace of $V_n(F)$.

A vector space V is said to be **spanned** or generated by the n -vectors X_1, X_2, \dots, X_m provided (a) the X_i lie in V and (b) every vector of V is a linear combination (11.1). Note that the vectors X_1, X_2, \dots, X_m are not restricted to be linearly independent.

Example 2. Let F be the field R of real numbers so that the 3-vectors $X_1 = [1, 1, 1]'$, $X_2 = [1, 2, 3]'$, $X_3 = [1, 3, 2]'$, and $X_4 = [3, 2, 1]'$ lie in ordinary space $S = V_3(R)$. Any vector $[a, b, c]'$ of S can be expressed as

$$y_1X_1 + y_2X_2 + y_3X_3 + y_4X_4 = \begin{bmatrix} y_1 + y_2 + y_3 + 3y_4 \\ y_1 + 2y_2 + 3y_3 + 2y_4 \\ y_1 + 3y_2 + 2y_3 + y_4 \end{bmatrix}$$

since the resulting system of equations

$$(1) \quad \begin{aligned} y_1 + y_2 + y_3 + 3y_4 &= a \\ y_1 + 2y_2 + 3y_3 + 2y_4 &= b \\ y_1 + 3y_2 + 2y_3 + y_4 &= c \end{aligned}$$

is consistent. Thus, the vectors X_1, X_2, X_3, X_4 span S .

The vectors X_1 and X_2 are linearly independent. They span a subspace (the plane π) of S which contains every vector $hX_1 + kX_2$, where h and k are real numbers.

The vector X_4 spans a subspace (the line L) of S which contains every vector hX_4 , where h is a real number.

See Problem 1.

BASIS AND DIMENSION. By the dimension of a vector space V is meant the maximum number of linearly independent vectors in V or, what is the same thing, the minimum number of linearly independent vectors required to span V . In elementary geometry, ordinary space is considered as a 3-space (space of dimension three) of points (a, b, c) . Here we have been considering it as a 3-space of vectors $[a, b, c]$. The plane π of Example 2 is of dimension 2 and the line L is of dimension 1.

A vector space of dimension r consisting of n -vectors will be denoted by $V_n^r(F)$. When $r = n$, we shall agree to write $V_n(F)$ for $V_n^n(F)$.

A set of r linearly independent vectors of $V_n^r(F)$ is called a **basis** of the space. Each vector of the space is then a **unique** linear combination of the vectors of this basis. All bases of $V_n^r(F)$ have exactly the same number of vectors but any r linearly independent vectors of the space will serve as a basis.

Example 3. The vectors X_1, X_2, X_3 of Example 2 span S since any vector $[a, b, c]$ of S can be expressed as

$$y_1X_1 + y_2X_2 + y_3X_3 = \begin{bmatrix} y_1 + y_2 + y_3 \\ y_1 + 2y_2 + 3y_3 \\ y_1 + 3y_2 + 2y_3 \end{bmatrix}$$

The resulting system of equations $\begin{cases} y_1 + y_2 + y_3 = a \\ y_1 + 2y_2 + 3y_3 = b \\ y_1 + 3y_2 + 2y_3 = c \end{cases}$ unlike the system (1), has a u-

nique solution. The vectors X_1, X_2, X_3 are a basis of S . The vectors X_1, X_2, X_4 are not a basis of S . (Show this.) They span the subspace π of Example 2, whose basis is the set X_1, X_2 .

Theorems I-V of Chapter 9 apply here, of course. In particular, Theorem IV may be restated as:

I. If X_1, X_2, \dots, X_m are a set of n -vectors over F and if r is the rank of the $n \times m$ matrix of their components, then from the set r linearly independent vectors may be selected. These r vectors span a $V_n^r(F)$ in which the remaining $m-r$ vectors lie.

See Problems 2-3.

Of considerable importance are:

II. If X_1, X_2, \dots, X_m are $m < n$ linearly independent n -vectors of $V_n(F)$ and if $X_{m+1}, X_{m+2}, \dots, X_n$ are any $n-m$ vectors of $V_n(F)$ which together with X_1, X_2, \dots, X_m form a linearly independent set, then the set X_1, X_2, \dots, X_n is a basis of $V_n(F)$.

See Problem 4.

III. If X_1, X_2, \dots, X_m are $m < n$ linearly independent n -vectors over F , then the p vectors

$$Y_j = \sum_{i=1}^m s_{ij} X_i \quad (j = 1, 2, \dots, p)$$

are linearly dependent if $p > m$ or, when $p \leq m$, if $[s_{ij}]$ is of rank $r < p$.

IV. If X_1, X_2, \dots, X_n are linearly independent n -vectors over F , then the vectors

$$Y_i = \sum_{j=1}^n a_{ij} X_j \quad (i = 1, 2, \dots, n)$$

are linearly independent if and only if $[a_{ij}]$ is nonsingular.

IDENTICAL SUBSPACES. If ${}_1V_n(F)$ and ${}_2V_n(F)$ are two subspaces of $V_n(F)$, they are identical if and only if each vector of ${}_1V_n(F)$ is a vector of ${}_2V_n(F)$ and conversely, that is, if and only if each is a subspace of the other.

See Problem 5.

SUM AND INTERSECTION OF TWO SPACES. Let $V_n^h(F)$ and $V_n^k(F)$ be two vector spaces. By their sum is meant the totality of vectors $X+Y$ where X is in $V_n^h(F)$ and Y is in $V_n^k(F)$. Clearly, this is a vector space; we call it the **sum space** $V_n^s(F)$. The dimension s of the sum space of two vector spaces does not exceed the sum of their dimensions.

By the **intersection** of the two vector spaces is meant the totality of vectors common to the two spaces. Now if X is a vector common to the two spaces, so also is aX ; likewise, if X and Y are common to the two spaces so also is $aX+bY$. Thus, the intersection of two spaces is a vector space; we call it the **intersection space** $V_n^t(F)$. The dimension of the intersection space of two vector spaces cannot exceed the smaller of the dimensions of the two spaces.

V. If two vector spaces $V_n^h(F)$ and $V_n^k(F)$ have $V_n^s(F)$ as sum space and $V_n^t(F)$ as intersection space, then $h+k = s+t$.

Example 4. Consider the subspace π_1 spanned by X_1 and X_2 of Example 2 and the subspace π_2 spanned by X_3 and X_4 . Since π_1 and π_2 are not identical (prove this) and since the four vectors span S , the sum space of π_1 and π_2 is S .

Now $4X_1 - X_2 = X_4$; thus, X_4 lies in both π_1 and π_2 . The subspace (line L) spanned by X_4 is then the intersection space of π_1 and π_2 . Note that π_1 and π_2 are each of dimension 2, S is of dimension 3, and L is of dimension 1. This agrees with Theorem V.

See Problems 6-8.

NULLITY OF A MATRIX. For a system of homogeneous equations $AX = 0$, the solution vectors X constitute a vector space called the **null space** of A . The dimension of this space, denoted by N_A , is called the **nullity** of A .

Restating Theorem VI, Chapter 10, we have

VI. If A has nullity N_A , then $AX = 0$ has N_A linearly independent solutions X_1, X_2, \dots

X_N such that every solution of $AX = 0$ is a linear combination of them and every such linear combination is a solution.

A basis for the null space of A is any set of N_A linearly independent solutions of $AX = 0$.

See Problem 9.

VII. For an $m \times n$ matrix A of rank r_A and nullity N_A ,

$$(11.2) \quad r_A + N_A = n$$

SYLVESTER'S LAWS OF NULLITY. If A and B are of order n and respective ranks r_A and r_B , the rank and nullity of their product AB satisfy the inequalities

$$(11.3) \quad \begin{aligned} r_{AB} &\geq r_A + r_B - n \\ N_{AB} &> N_A, \quad N_{AB} > N_B \\ N_{AB} &\leq N_A + N_B \end{aligned}$$

See Problem 10.

BASES AND COORDINATES. The n -vectors

$$E_1 = [1, 0, 0, \dots, 0]', \quad E_2 = [0, 1, 0, \dots, 0]', \quad \dots, \quad E_n = [0, 0, 0, \dots, 1]'$$

are called **elementary** or **unit vectors** over F . The elementary vector E_j , whose j th component is 1, is called the j th elementary vector. The elementary vectors E_1, E_2, \dots, E_n constitute an important basis for $V_n(F)$.

Every vector $X = [x_1, x_2, \dots, x_n]'$ of $V_n(F)$ can be expressed uniquely as the sum

$$X = \sum_{i=1}^n x_i E_i = x_1 E_1 + x_2 E_2 + \dots + x_n E_n$$

of the elementary vectors. The components x_1, x_2, \dots, x_n of X are now called the **coordinates** of X relative to the E -basis. Hereafter, unless otherwise specified, we shall assume that a vector X is given relative to this basis.

Let Z_1, Z_2, \dots, Z_n be another basis of $V_n(F)$. Then there exist unique scalars a_1, a_2, \dots, a_n in F such that

$$X = \sum_{i=1}^n a_i Z_i = a_1 Z_1 + a_2 Z_2 + \dots + a_n Z_n$$

These scalars a_1, a_2, \dots, a_n are called the coordinates of X relative to the Z -basis. Writing $X_Z = [a_1, a_2, \dots, a_n]'$, we have

$$(11.4) \quad X = [Z_1, Z_2, \dots, Z_n] X_Z = Z \cdot X_Z$$

where Z is the matrix whose columns are the basis vectors Z_1, Z_2, \dots, Z_n .

Example 5. If $Z_1 = [2, -1, 3]'$, $Z_2 = [1, 2, -1]'$, $Z_3 = [1, -1, -1]'$ is a basis of $V_3(F)$ and $X_Z = [1, 2, 3]'$ is a vector of $V_3(F)$ relative to that basis, then

$$X = [Z_1, Z_2, Z_3] X_Z = \begin{bmatrix} 2 & 1 & 1 \\ -1 & 2 & -1 \\ 3 & -1 & -1 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} = \begin{bmatrix} 7 \\ 0 \\ -2 \end{bmatrix} = [7, 0, -2]'$$

relative to the E -basis.

See Problem 11.

$$X_E = (7, 0, 2)'$$

↑
coordinates of X in E -basis

Let W_1, W_2, \dots, W_n be yet another basis of $V_n(F)$. Suppose $X_W = [b_1, b_2, \dots, b_n]'$ so that

$$(11.5) \quad X = [W_1, W_2, \dots, W_n] X_W = W \cdot X_W$$

From (11.4) and (11.5), $X = Z \cdot X_Z = W \cdot X_W$ and

$$(11.6) \quad X_W = W^{-1} \cdot Z \cdot X_Z = P X_Z$$

where $P = W^{-1}Z$.

Thus,

VIII. If a vector of $V_n(F)$ has coordinates X_Z and X_W respectively relative to two bases of $V_n(F)$, then there exists a non-singular matrix P , determined solely by the two bases and given by (11.6) such that $X_W = P X_Z$.

See Problem 12.

SOLVED PROBLEMS

1. The set of all vectors $X = [x_1, x_2, x_3, x_4]'$, where $x_1 + x_2 + x_3 + x_4 = 0$ is a subspace V of $V_4(F)$ since the sum of any two vectors of the set and any scalar multiple of a vector of the set have components whose sum is zero, that is, are vectors of the set.

2. Since $\begin{bmatrix} 1 & 3 & 1 \\ 2 & 4 & 0 \\ 2 & 4 & 0 \\ 1 & 3 & 1 \end{bmatrix}$ is of rank 2, the vectors $X_1 = [1, 2, 2, 1]'$, $X_2 = [3, 4, 4, 3]'$, and $X_3 = [1, 0, 0, 1]'$ are linearly dependent and span a vector space $V_4^2(F)$.

Now any two of these vectors are linearly independent; hence, we may take X_1 and X_2 , X_1 and X_3 , or X_2 and X_3 as a basis of the $V_4^2(F)$.

3. Since $\begin{bmatrix} 1 & 4 & 2 & 4 \\ 1 & 3 & 1 & 2 \\ 1 & 2 & 0 & 0 \\ 0 & -1 & -1 & -2 \end{bmatrix}$ is of rank 2, the vectors $X_1 = [1, 1, 1, 0]'$, $X_2 = [4, 3, 2, -1]'$, $X_3 = [2, 1, 0, -1]'$, and $X_4 = [4, 2, 0, -2]'$ are linearly dependent and span a $V_4^2(F)$.

For a basis, we may take any two of the vectors except the pair X_3, X_4 .

4. The vectors X_1, X_2, X_3 of Problem 2 lie in $V_4(F)$. Find a basis.

For a basis of this space we may take $X_1, X_2, X_4 = [1, 0, 0, 0]'$, and $X_5 = [0, 1, 0, 0]'$ or $X_1, X_2, X_6 = [1, 2, 3, 4]'$, and $X_7 = [1, 3, 6, 8]'$, since the matrices $[X_1, X_2, X_4, X_5]$ and $[X_1, X_2, X_6, X_7]$ are of rank 4.

5. Let $X_1 = [1, 2, 1]'$, $X_2 = [1, 2, 3]'$, $X_3 = [3, 6, 5]'$, $Y_1 = [0, 0, 1]'$, $Y_2 = [1, 2, 5]'$ be vectors of $V_3(F)$. Show that the space spanned by X_1, X_2, X_3 and the space spanned by Y_1, Y_2 are identical.

First, we note that X_1 and X_2 are linearly independent while $X_3 = 2X_1 + X_2$. Thus, the X_i span a space of dimension two, say ${}_1V_3^2(F)$. Also, the Y_i being linearly independent span a space of dimension two, say ${}_2V_3^2(F)$.

Next, $Y_1 = \frac{1}{2}X_2 - \frac{1}{2}X_1$, $Y_2 = 2X_2 - X_1$; $X_1 = Y_2 - 4Y_1$, $X_2 = Y_2 - 2Y_1$. Thus, any vector $aY_1 + bY_2$ of ${}_2V_3^2(F)$ is a vector $(\frac{1}{2}a + 2b)X_2 - (\frac{1}{2}a + b)X_1$ of ${}_1V_3^2(F)$ and any vector $cX_1 + dX_2$ of ${}_1V_3^2(F)$ is a vector $(c+d)Y_2 - (4c+2d)Y_1$ of ${}_2V_3^2(F)$. Hence, the two spaces are identical.

6. (a) If $X = [x_1, x_2, x_3]'$ lies in the $V_3^2(F)$ spanned by $X_1 = [1, -1, 1]'$ and $X_2 = [3, 4, -2]'$, then

$$\begin{vmatrix} x_1 & 1 & 3 \\ x_2 & -1 & 4 \\ x_3 & 1 & -2 \end{vmatrix} = -2x_1 + 5x_2 + 7x_3 = 0.$$

- (b) If $X = [x_1, x_2, x_3, x_4]'$ lies in the $V_4^2(F)$ spanned by $X_1 = [1, 1, 2, 3]'$ and $X_2 = [1, 0, -2, 1]'$, then

$$\begin{vmatrix} x_1 & 1 & 1 \\ x_2 & 1 & 0 \\ x_3 & 2 & -2 \\ x_4 & 3 & 1 \end{vmatrix} \text{ is of rank 2. Since } \begin{vmatrix} 1 & 1 \\ 1 & 0 \end{vmatrix} \neq 0, \text{ this requires } \begin{vmatrix} x_1 & 1 & 1 \\ x_2 & 1 & 0 \\ x_3 & 2 & -2 \end{vmatrix} = -2x_1 + 4x_2 - x_3 = 0 \text{ and}$$

$$\begin{vmatrix} x_1 & 1 & 1 \\ x_2 & 1 & 0 \\ x_4 & 3 & 1 \end{vmatrix} = x_1 + 2x_2 - x_4 = 0.$$

These problems illustrate that every $V_n^k(F)$ may be defined as the totality of solutions over F of a system of $n-k$ linearly independent homogeneous linear equations over F in n unknowns.

7. Prove: If two vector spaces $V_n^h(F)$ and $V_n^k(F)$ have $V_n^s(F)$ as sum space and $V_n^t(F)$ as intersection space, then $h+k = s+t$.

Suppose $t=h$; then $V_n^h(F)$ is a subspace of $V_n^k(F)$ and their sum space is V_n^k itself. Thus, $s=k$, $t=h$ and $s+t = h+k$. The reader will show that the same is true if $t=k$.

Suppose next that $t < h$, $t < k$ and let X_1, X_2, \dots, X_t span $V_n^t(F)$. Then by Theorem II there exist vectors $Y_{t+1}, Y_{t+2}, \dots, Y_h$ so that $X_1, X_2, \dots, X_t, Y_{t+1}, \dots, Y_h$ span $V_n^h(F)$ and vectors $Z_{t+1}, Z_{t+2}, \dots, Z_k$ so that $X_1, X_2, \dots, X_t, Z_{t+1}, \dots, Z_k$ span $V_n^k(F)$.

Now suppose there exist scalars a 's and b 's such that

$$(11.4) \quad \sum_{i=1}^t a_i X_i + \sum_{i=t+1}^h a_i Y_i + \sum_{i=t+1}^k b_i Z_i = 0 \quad \text{or}$$

$$\sum_{i=1}^t a_i X_i + \sum_{i=t+1}^h a_i Y_i = - \sum_{i=t+1}^k b_i Z_i$$

The vector on the left belongs to $V_n^h(F)$, and from the right member, belongs also to $V_n^k(F)$; thus it belongs to $V_n^t(F)$. But X_1, X_2, \dots, X_t span $V_n^t(F)$; hence, $a_{t+1} = a_{t+2} = \dots = a_h = 0$.

$$\text{Now from (11.4),} \quad \sum_{i=1}^t a_i X_i + \sum_{i=t+1}^k b_i Z_i = 0$$

But the X 's and Z 's are linearly independent so that $a_1 = a_2 = \dots = a_t = b_{t+1} = b_{t+2} = \dots = b_k = 0$; thus, the X 's, Y 's, and Z 's are a linearly independent set and span $V_n^s(F)$. Then $s = h+k-t$ as was to be proved.

8. Consider ${}_1V_3^2(F)$ having $X_1 = [1, 2, 3]'$ and $X_2 = [1, 1, 1]'$ as basis and ${}_2V_3^2(F)$ having $Y_1 = [3, 1, 2]'$

and $Y_2 = [1, 0, 1]'$ as basis. Since the matrix of the components $\begin{bmatrix} 1 & 1 & 3 & 1 \\ 2 & 1 & 1 & 0 \\ 3 & 1 & 2 & 1 \end{bmatrix}$ is of rank 3, the sum

space is $V_3(F)$. As a basis, we may take $X_1, X_2,$ and Y_1 .

From $h + k = s + t$, the intersection space is a $V_3^1(F)$. To find a basis, we equate linear combinations of the vectors of the bases of ${}_1V_3^2(F)$ and ${}_2V_3^2(F)$ as

$$aX_1 + bX_2 = cY_1 + dY_2$$

take $d = 1$ for convenience, and solve $\begin{cases} a + b - 3c = 1 \\ 2a + b - c = 0 \\ 3a + b - 2c = 1 \end{cases}$ obtaining $a = 1/3, b = -4/3, c = -2/3$. Then

$aX_1 + bX_2 = [-1, -2/3, -1/3]'$ is a basis of the intersection space. The vector $[3, 2, 1]'$ is also a basis.

9. Determine a basis for the null space of $A = \begin{bmatrix} 1 & 1 & 3 & 3 \\ 0 & 2 & 2 & 4 \\ 1 & 0 & 2 & 1 \\ 1 & 1 & 3 & 3 \end{bmatrix}$

Consider the system of equations $AX = 0$ which reduces to $\begin{cases} x_1 + 2x_3 + x_4 = 0 \\ x_2 + x_3 + 2x_4 = 0 \end{cases}$

A basis for the null space of A is the pair of linearly independent solutions $[1, 2, 0, -1]'$ and $[2, 1, -1, 0]'$ of these equations.

10. Prove: $r_{AB} \geq r_A + r_B - n$.

Suppose first that A has the form $\begin{bmatrix} I_{r_A} & 0 \\ 0 & 0 \end{bmatrix}$. Then the first r_A rows of AB are the first r_A rows of B while the remaining rows are zeros. By Problem 10, Chapter 5, the rank of AB is $r_{AB} \geq r_A + r_B - n$.

Suppose next that A is not of the above form. Then there exist nonsingular matrices P and Q such that PAQ has that form while the rank of $PAQB$ is exactly that of AB (why?).

The reader may consider the special case when $B = \begin{bmatrix} I_{r_B} & 0 \\ 0 & 0 \end{bmatrix}$.

11. Let $X = [1, 2, 1]'$ relative to the E -basis. Find its coordinates relative to a new basis $Z_1 = [1, 1, 0]'$, $Z_2 = [1, 0, 1]'$, and $Z_3 = [1, 1, 1]'$.

Solution (a). Write

$$(i) X = aZ_1 + bZ_2 + cZ_3, \text{ that is, } \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix} = a \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} + b \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} + c \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}. \text{ Then } \begin{cases} a + b + c = 1 \\ a + c = 2 \text{ and } a = 0, b = -1, \\ b + c = 1 \end{cases}$$

$c = 2$. Thus relative to the Z -basis, we have $X_Z = [0, -1, 2]'$.

Solution (b). Rewriting (i) as $X = [Z_1, Z_2, Z_3]X_Z = ZX_Z$, we have

$$X_Z = Z^{-1}X = \begin{bmatrix} 1 & 0 & -1 \\ 1 & -1 & 0 \\ -1 & 1 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix} = [0, -1, 2]'$$

12. Let X_Z and X_W be the coordinates of a vector X with respect to the two bases $Z_1 = [1, 1, 0]'$, $Z_2 = [1, 0, 1]'$, $Z_3 = [1, 1, 1]'$ and $W_1 = [1, 1, 2]'$, $W_2 = [2, 2, 1]'$, $W_3 = [1, 2, 2]'$. Determine the matrix P such that $X_W = PX_Z$.

$$\text{Here } Z = [Z_1, Z_2, Z_3] = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 0 & 1 \\ 0 & 1 & 1 \end{bmatrix}, \quad W = \begin{bmatrix} 1 & 2 & 1 \\ 1 & 2 & 2 \\ 2 & 1 & 2 \end{bmatrix}, \quad \text{and } W^{-1} = \frac{1}{3} \begin{bmatrix} 2 & -3 & 2 \\ 2 & 0 & -1 \\ -3 & 3 & 0 \end{bmatrix}$$

$$\text{Then } P = W^{-1}Z = \frac{1}{3} \begin{bmatrix} -1 & 4 & 1 \\ 2 & 1 & 1 \\ 0 & -3 & 0 \end{bmatrix} \text{ by (11.6).}$$

SUPPLEMENTARY PROBLEMS

13. Let $[x_1, x_2, x_3, x_4]'$ be an arbitrary vector of $V_4(R)$, where R denotes the field of real numbers. Which of the following sets are subspaces of $V_4(R)$?

(a) All vectors with $x_1 = x_2 = x_3 = x_4$.

(d) All vectors with $x_1 = 1$.

(b) All vectors with $x_1 = x_2$, $x_3 = 2x_4$.

(e) All vectors with x_1, x_2, x_3, x_4 integral.

(c) All vectors with $x_4 = 0$.

Ans. All except (d) and (e).

14. Show that $[1, 1, 1, 1]'$ and $[2, 3, 3, 2]'$ are a basis of the $V_4(F)$ of Problem 2.

15. Determine the dimension of the vector space spanned by each set of vectors. Select a basis for each.

(a) $[1, 2, 3, 4, 5]'$
 $[5, 4, 3, 2, 1]'$
 $[1, 1, 1, 1, 1]'$

(b) $[1, 1, 0, -1]'$
 $[1, 2, 3, 4]'$
 $[2, 3, 3, 3]'$

(c) $[1, 1, 1, 1, 1]'$
 $[3, 4, 5, 6]'$
 $[1, 2, 3, 4]'$
 $[1, 0, -1, -2]'$

Ans. (a), (b), (c), $r = 2$

16. (a) Show that the vectors $X_1 = [1, -1, 1]'$ and $X_2 = [3, 4, -2]'$ span the same space as $Y_1 = [9, 5, -1]'$ and $Y_2 = [-17, -11, 3]'$.

(b) Show that the vectors $X_1 = [1, -1, 1]'$ and $X_2 = [3, 4, -2]'$ do not span the same space as $Y_1 = [-2, 2, -2]'$ and $Y_2 = [4, 3, 1]'$.

17. Show that if the set X_1, X_2, \dots, X_k is a basis for $V_n^k(F)$, then any other vector Y of the space can be represented *uniquely* as a linear combination of X_1, X_2, \dots, X_k .

Hint. Assume $Y = \sum_{i=1}^k a_i X_i = \sum_{i=1}^k b_i X_i$.

18. Consider the 4×4 matrix whose columns are the vectors of a basis of the $V_4^2(R)$ of Problem 2 and a basis of the $V_4^2(R)$ of Problem 3. Show that the rank of this matrix is 4; hence, $V_4(R)$ is the sum space and $V_4^0(R)$, the zero space, is the intersection space of the two given spaces.

19. Follow the proof given in Problem 8, Chapter 10, to prove Theorem III.

20. Show that the space spanned by $[1.0.0.0.0]'$, $[0.0.0.0.1]'$, $[1.0.1.0.0]'$, $[0.0.1.0.0]'$, $[1.0.0.1.1]'$ and the space spanned by $[1.0.0.0.1]'$, $[0.1.0.1.0]'$, $[0.1.-2.1.0]'$, $[1.0.-1.0.1]'$, $[0.1.1.1.0]'$ are of dimensions 4 and 3, respectively. Show that $[1.0.1.0.1]'$ and $[1.0.2.0.1]'$ are a basis for the intersection space.

21. Find, relative to the basis $Z_1 = [1.1.2]'$, $Z_2 = [2.2.1]'$, $Z_3 = [1.2.2]'$ the coordinates of the vectors (a) $[1.1.0]'$, (b) $[1.0.1]'$, (c) $[1.1.1]'$.

Ans. (a) $[-1/3, 2/3, 0]'$, (b) $[4/3, 1/3, -1]'$, (c) $[1/3, 1/3, 0]'$

22. Find, relative to the basis $Z_1 = [0.1.0]'$, $Z_2 = [1.1.1]'$, $Z_3 = [3.2.1]'$ the coordinates of the vectors

(a) $[2.-1.0]'$, (b) $[1.-3.5]'$, (c) $[0.0.1]'$.

Ans. (a) $[-2, -1.1]'$, (b) $[-6.7, -2]'$, (c) $[-1/2, 3/2, -1/2]'$

23. Let X_Z and X_W be the coordinates of a vector X with respect to the given pair of bases. Determine the matrix P such that $X_W = PX_Z$.

(a) $Z_1 = [1.0.0]'$, $Z_2 = [1.0.1]'$, $Z_3 = [1.1.1]'$

$W_1 = [0.1.0]'$, $W_2 = [1.2.3]'$, $W_3 = [1.-1.1]'$

(b) $Z_1 = [0.1.0]'$, $Z_2 = [1.1.0]'$, $Z_3 = [1.2.3]'$

$W_1 = [1.1.0]'$, $W_2 = [1.1.1]'$, $W_3 = [1.2.1]'$

Ans. (a) $P = \frac{1}{2} \begin{bmatrix} 5 & 2 & 4 \\ -1 & 0 & 0 \\ 3 & 2 & 2 \end{bmatrix}$, (b) $P = \begin{bmatrix} 0 & 1 & -2 \\ -1 & 0 & 2 \\ 1 & 0 & 1 \end{bmatrix}$

24. Prove: If P_j is a solution of $AX = E_j$, ($j = 1, 2, \dots, n$), then $\sum_{j=1}^n h_j P_j$ is a solution of $AX = H$, where $H = [h_1, h_2, \dots, h_n]'$.

Hint. $H = h_1 E_1 + h_2 E_2 + \dots + h_n E_n$.

25. The vector space defined by all linear combinations of the columns of a matrix A is called the **column space** of A . The vector space defined by all linear combinations of the rows of A is called the **row space** of A . Show that the columns of AB are in the column space of A and the rows of AB are in the row space of B .

26. Show that $AX = H$, a system of m non-homogeneous equations in n unknowns, is consistent if and only if the vector H belongs to the column space of A .

27. Determine a basis for the null space of (a) $\begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & -1 \\ 1 & 0 & 1 \end{bmatrix}$, (b) $\begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 2 & 1 & 2 \\ 3 & 4 & 3 & 4 \end{bmatrix}$.

Ans. (a) $[1, -1, -1]'$, (b) $[1, 1, -1, -1]'$, $[1, 2, -1, -2]'$

28. Prove: (a) $N_{AB} \supseteq N_A \cdot N_{AB} \supseteq N_B$ (b) $N_{AB} \subseteq N_A + N_B$

Hint: (a) $N_{AB} = n - r_{AB}$; $r_{AB} \leq r_A$ and r_B .

(b) Consider $n - r_{AB}$, using the theorem of Problem 10.

29. Derive a procedure for Problem 16 using only column transformations on $A = [X_1, X_2, Y_1, Y_2]$. Then resolve Problem 5.

Chapter 12

Linear Transformations

DEFINITION. Let $X = [x_1, x_2, \dots, x_n]'$ and $Y = [y_1, y_2, \dots, y_n]'$ be two vectors of $V_n(F)$, their coordinates being relative to the same basis of the space. Suppose that the coordinates of X and Y are related by

$$(12.1) \quad \begin{cases} y_1 = a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n \\ y_2 = a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n \\ \dots \\ y_n = a_{n1}x_1 + a_{n2}x_2 + \dots + a_{nn}x_n \end{cases}$$

or, briefly, $Y = AX$

where $A = [a_{ij}]$ is over F . Then (12.1) is a transformation T which carries any vector X of $V_n(F)$ into (usually) another vector Y of the same space, called its image.

If (12.1) carries X_1 into Y_1 and X_2 into Y_2 , then

- (a) it carries kX_1 into kY_1 , for every scalar k , and
- (b) it carries $aX_1 + bX_2$ into $aY_1 + bY_2$, for every pair of scalars a and b . For this reason, the transformation is called **linear**.

Example 1. Consider the linear transformation $Y = AX = \begin{bmatrix} 1 & 1 & 2 \\ 1 & 2 & 5 \\ 1 & 3 & 3 \end{bmatrix} X$ in ordinary space $V_3(R)$.

(a) The image of $X = [2, 0, 5]'$ is $Y = \begin{bmatrix} 1 & 1 & 2 \\ 1 & 2 & 5 \\ 1 & 3 & 3 \end{bmatrix} \begin{bmatrix} 2 \\ 0 \\ 5 \end{bmatrix} = \begin{bmatrix} 12 \\ 27 \\ 17 \end{bmatrix} = [12, 27, 17]'$.

(b) The vector X whose image is $Y = [2, 0, 5]'$ is obtained by solving $\begin{bmatrix} 1 & 1 & 2 \\ 1 & 2 & 5 \\ 1 & 3 & 3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 2 \\ 0 \\ 5 \end{bmatrix}$.

Since $\begin{bmatrix} 1 & 1 & 2 & 2 \\ 1 & 2 & 5 & 0 \\ 1 & 3 & 3 & 5 \end{bmatrix} \sim \begin{bmatrix} 1 & 0 & 0 & 13/5 \\ 0 & 1 & 0 & 11/5 \\ 0 & 0 & 1 & -7/5 \end{bmatrix}$, $X = [13/5, 11/5, -7/5]'$.

BASIC THEOREMS. If in (12.1), $X = [1, 0, \dots, 0] = E_1$ then $Y = [a_{11}, a_{21}, \dots, a_{n1}]'$ and, in general, if $X = E_j$ then $Y = [a_{1j}, a_{2j}, \dots, a_{nj}]'$. Hence,

I. A linear transformation (12.1) is uniquely determined when the images (Y 's) of the basis vectors are known, the respective columns of A being the coordinates of the images of these vectors. See Problem 1.

A linear transformation (12.1) is called **non-singular** if the images of distinct vectors X_i are distinct vectors Y_i . Otherwise the transformation is called **singular**.

II. A linear transformation (12.1) is non-singular if and only if A , the matrix of the transformation, is non-singular. See Problem 2.

III. A non-singular linear transformation carries linearly independent (dependent) vectors into linearly independent (dependent) vectors. See Problem 3.

From Theorem III follows

IV. Under a non-singular transformation (12.1) the image of a vector space $V_n^k(F)$ is a vector space $V_n^k(F)$, that is, the dimension of the vector space is preserved. In particular, the transformation is a mapping of $V_n(F)$ onto itself.

When A is non-singular, the inverse of (12.1)

$$X = A^{-1}Y$$

carries the set of vectors Y_1, Y_2, \dots, Y_n whose components are the columns of A into the basis vectors of the space. It is also a linear transformation.

V. The elementary vectors E_i of $V_n(F)$ may be transformed into any set of n linearly independent n -vectors by a non-singular linear transformation and conversely.

VI. If $Y = AX$ carries a vector X into a vector Y , if $Z = BY$ carries Y into Z , and if $W = CZ$ carries Z into W , then $Z = BY = (BA)X$ carries X into Z and $W = (CBA)X$ carries X into W .

VII. When any two sets of n linearly independent n -vectors are given, there exists a non-singular linear transformation which carries the vectors of one set into the vectors of the other.

CHANGE OF BASIS. Relative to a Z -basis, let $Y_Z = AX_Z$ be a linear transformation of $V_n(F)$. Suppose that the basis is changed and let X_W and Y_W be the coordinates of X_Z and Y_Z respectively relative to the new basis. By Theorem VIII, Chapter 11, there exists a non-singular matrix P such that $X_W = PX_Z$ and $Y_W = PY_Z$ or, setting $P^{-1} = Q$, such that

$$X_Z = QX_W \quad \text{and} \quad Y_Z = QY_W$$

Then
$$Y_W = Q^{-1}Y_Z = Q^{-1}AX_Z = Q^{-1}AQX_W = BX_W$$

where

(12.2)
$$B = Q^{-1}AQ$$

Two matrices A and B such that there exists a non-singular matrix Q for which $B = Q^{-1}AQ$ are called **similar**. We have proved

VIII. If $Y_Z = AX_Z$ is a linear transformation of $V_n(F)$ relative to a given basis (Z -basis) and $Y_W = BX_W$ is the same linear transformation relative to another basis (W -basis), then A and B are similar.

Note. Since $Q = P^{-1}$, (12.2) might have been written as $B = PAP^{-1}$. A study of similar matrices will be made later. There we shall agree to write $B = R^{-1}AR$ instead of $B = SAS^{-1}$ but for no compelling reason.

Example 2. Let $Y = AX = \begin{bmatrix} 1 & 1 & 3 \\ 1 & 2 & 1 \\ 1 & 3 & 2 \end{bmatrix} X$ be a linear transformation relative to the E -basis and let $W_1 =$

$[1, 2, 1]'$, $W_2 = [1, -1, 2]'$, $W_3 = [1, -1, -1]'$ be a new basis. (a) Given the vector $X = [3, 0, 2]'$, find the coordinates of its image relative to the W -basis. (b) Find the linear transformation $Y_W = BX_W$ corresponding to $Y = AX$. (c) Use the result of (b) to find the image Y_W of $X_W = [1, 3, 3]'$.

$$\text{Write } W = [W_1, W_2, W_3] = \begin{bmatrix} 1 & 1 & 1 \\ 2 & -1 & -1 \\ 1 & 2 & -1 \end{bmatrix}; \text{ then } W^{-1} = \frac{1}{9} \begin{bmatrix} 3 & 3 & 0 \\ 1 & -2 & 3 \\ 5 & -1 & -3 \end{bmatrix}.$$

(a) Relative to the W -basis, the vector $X = [3, 0, 2]'$ has coordinates $X_W = W^{-1}X = [1, 1, 1]'$. The image of X is $Y = AX = [9, 5, 7]'$ which, relative to the W -basis is $Y_W = W^{-1}Y = [14/3, 20/9, 19/9]'$.

$$(b) Y_W = W^{-1}Y = W^{-1}AX = (W^{-1}AW)X_W = BX_W = \frac{1}{9} \begin{bmatrix} 36 & 21 & -15 \\ 21 & 10 & -11 \\ -3 & 23 & -1 \end{bmatrix} X_W$$

$$(c) Y_W = \frac{1}{9} \begin{bmatrix} 36 & 21 & -15 \\ 21 & 10 & -11 \\ -3 & 23 & -1 \end{bmatrix} \begin{bmatrix} 1 \\ 3 \\ 3 \end{bmatrix} = \begin{bmatrix} 6 \\ 2 \\ -7 \end{bmatrix} = [6, 2, 7]'$$

See Problem 5.

SOLVED PROBLEMS

- Set up the linear transformation $Y = AX$ which carries E_1 into $Y_1 = [1, 2, 3]'$, E_2 into $[3, 1, 2]'$, and E_3 into $Y_3 = [2, 1, 3]'$.
 - Find the images of $X_1 = [1, 1, 1]'$, $X_2 = [3, -1, 4]'$, and $X_3 = [4, 0, 5]'$.
 - Show that X_1 and X_2 are linearly independent as also are their images.
 - Show that $X_1, X_2,$ and X_3 are linearly dependent as also are their images.

(a) By Theorem I, $A = [Y_1, Y_2, Y_3]$; the equation of the linear transformation is $Y = AX = \begin{bmatrix} 1 & 3 & 2 \\ 2 & 1 & 1 \\ 3 & 2 & 3 \end{bmatrix} X$.

(b) The image of $X_1 = [1, 1, 1]'$ is $Y_1 = \begin{bmatrix} 1 & 3 & 2 \\ 2 & 1 & 1 \\ 3 & 2 & 3 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} = [6, 4, 8]'$. The image of X_2 is $Y_2 = [8, 9, 19]'$ and the image of X_3 is $Y_3 = [14, 13, 27]'$.

(c) The rank of $[X_1, X_2] = \begin{bmatrix} 1 & 3 \\ 1 & -1 \\ 1 & 4 \end{bmatrix}$ is 2 as also is that of $[Y_1, Y_2] = \begin{bmatrix} 6 & 8 \\ 4 & 9 \\ 8 & 19 \end{bmatrix}$. Thus, X_1 and X_2 are linearly independent as also are their images.

(d) We may compare the ranks of $[X_1, X_2, X_3]$ and $[Y_1, Y_2, Y_3]$; however, $X_3 = X_1 + X_2$ and $Y_3 = Y_1 + Y_2$ so that both sets are linearly dependent.

2. Prove: A linear transformation (12.1) is non-singular if and only if A is non-singular.

Suppose A is non-singular and the transforms of $X_1 \neq X_2$ are $Y = AX_1 = AX_2$. Then $A(X_1 - X_2) = 0$ and the system of homogeneous linear equations $AX = 0$ has the non-trivial solution $X = X_1 - X_2$. This is possible if and only if $|A| = 0$, a contradiction of the hypothesis that A is non-singular.

3. Prove: A non-singular linear transformation carries linearly independent vectors into linearly independent vectors.

Assume the contrary, that is, suppose that the images $Y_i = AX_i$, ($i = 1, 2, \dots, p$) of the linearly independent vectors X_1, X_2, \dots, X_p are linearly dependent. Then there exist scalars s_1, s_2, \dots, s_p , not all zero, such that

$$\sum_{i=1}^p s_i Y_i = s_1 Y_1 + s_2 Y_2 + \dots + s_p Y_p = 0$$

or

$$\sum_{i=1}^p s_i (AX_i) = A(s_1 X_1 + s_2 X_2 + \dots + s_p X_p) = 0$$

Since A is non-singular, $s_1 X_1 + s_2 X_2 + \dots + s_p X_p = 0$. But this is contrary to the hypothesis that the X_i are linearly independent. Hence, the Y_i are linearly independent.

4. A certain linear transformation $Y = AX$ carries $X_1 = [1, 0, 1]'$ into $[2, 3, -1]'$, $X_2 = [1, -1, 1]'$ into $[3, 0, -2]'$, and $X_3 = [1, 2, -1]'$ into $[-2, 7, -1]'$. Find the images of E_1, E_2, E_3 and write the equation of the transformation.

Let $aX_1 + bX_2 + cX_3 = E_1$; then

$$\begin{cases} a + b + c = 1 \\ -b + 2c = 0 \\ a + b - c = 0 \end{cases} \text{ and } a = -\frac{1}{2}, b = 1, c = \frac{1}{2}. \text{ Thus, } E_1 = -\frac{1}{2}X_1 + X_2 + \frac{1}{2}X_3$$

and its image is $Y_1 = -\frac{1}{2}[2, 3, -1]' + [3, 0, -2]' + \frac{1}{2}[-2, 7, -1]' = [1, 2, -2]'$. Similarly, the image of E_2 is $Y_2 = [-1, 3, 1]'$ and the image of E_3 is $Y_3 = [1, 1, 1]'$. The equation of the transformation is

$$Y = [Y_1, Y_2, Y_3]X = \begin{bmatrix} 1 & -1 & 1 \\ 2 & 3 & 1 \\ -2 & 1 & 1 \end{bmatrix} X$$

5. If $Y_Z = AX_Z = \begin{bmatrix} 1 & 1 & 2 \\ 2 & 2 & 1 \\ 3 & 1 & 2 \end{bmatrix} X_Z$ is a linear transformation relative to the Z -basis of Problem 12, Chapter 11, find the same transformation $Y_W = BX_W$ relative to the W -basis of that problem.

From Problem 12, Chapter 11, $X_W = PX_Z = \frac{1}{3} \begin{bmatrix} -1 & 4 & 1 \\ 2 & 1 & 1 \\ 0 & -3 & 0 \end{bmatrix} X_Z$. Then

$$X_Z = P^{-1}X_W = \begin{bmatrix} -1 & 1 & -1 \\ 0 & 0 & -1 \\ 2 & 1 & 3 \end{bmatrix} X_W = QX_W$$

and

$$Y_W = PY_Z = Q^{-1}AX_Z = Q^{-1}AQX_W = \frac{1}{3} \begin{bmatrix} -2 & 14 & -6 \\ 7 & 14 & 9 \\ 0 & -9 & 3 \end{bmatrix} X_W$$

SUPPLEMENTARY PROBLEMS

6. In Problem 1 show: (a) the transformation is non-singular, (b) $X = A^{-1}Y$ carries the column vectors of A into the elementary vectors.

7. Using the transformation of Problem 1, find (a) the image of $X = [1, 1, 2]'$, (b) the vector X whose image is $[-2, -5, -5]'$. *Ans.* (a) $[8, 5, 11]'$, (b) $[-3, -1, 2]'$

8. Study the effect of the transformation $Y = IX$, also $Y = kIX$.

9. Set up the linear transformation which carries E_1 into $[1, 2, 3]'$, E_2 into $[3, 1, 2]'$, and E_3 into $[2, -1, -1]'$. Show that the transformation is singular and carries the linearly independent vectors $[1, 1, 1]'$ and $[2, 0, 2]'$ into the same image vector.

10. Suppose (12.1) is non-singular and show that if X_1, X_2, \dots, X_n are linearly dependent so also are their images Y_1, Y_2, \dots, Y_n .

11. Use Theorem III to show that under a non-singular transformation the dimension of a vector space is unchanged. *Hint.* Consider the images of a basis of $V_n^k(F)$.

12. Given the linear transformation $Y = \begin{bmatrix} 1 & 1 & 0 \\ 2 & 3 & 1 \\ -2 & 3 & 5 \end{bmatrix} X$, show (a) it is singular, (b) the images of the linearly independent vectors $X_1 = [1, 1, 1]'$, $X_2 = [2, 1, 2]'$, and $X_3 = [1, 2, 3]'$ are linearly dependent, (c) the image of $V_3(R)$ is a $V_3^2(R)$.

13. Given the linear transformation $Y = \begin{bmatrix} 1 & 1 & 3 \\ 1 & 2 & 4 \\ 1 & 1 & 3 \end{bmatrix} X$, show (a) it is singular, (b) the image of every vector of the $V_3^2(R)$ spanned by $[1, 1, 1]'$ and $[3, 2, 0]'$ lies in the $V_3^1(R)$ spanned by $[5, 7, 5]'$.

14. Prove Theorem VII.

Hint. Let X_i and Y_i , ($i=1, 2, \dots, n$) be the given sets of vectors. Let $Z = AX$ carry the set X_i into E_i and $Y = BZ$ carry the E_i into Y_i .

15. Prove: Similar matrices have equal determinants.

16. Let $Y = AX = \begin{bmatrix} 1 & 2 & 3 \\ 3 & 2 & 1 \\ 1 & 1 & 1 \end{bmatrix} X$ be a linear transformation relative to the E -basis and let a new basis, say $Z_1 =$

$[1, 1, 0]'$, $Z_2 = [1, 0, 1]'$, $Z_3 = [1, 1, 1]'$ be chosen. Let $X = [1, 2, 3]'$ relative to the E -basis. Show that

(a) $Y = [14, 10, 6]'$ is the image of X under the transformation.

(b) X , when referred to the new basis, has coordinates $X_Z = [-2, -1, 4]'$ and Y has coordinates $Y_Z = [8, 4, 2]'$.

(c) $X_Z = PX$ and $Y_Z = PY$, where $P = \begin{bmatrix} 1 & 0 & -1 \\ 1 & -1 & 0 \\ -1 & 1 & 1 \end{bmatrix} = [Z_1, Z_2, Z_3]^{-1}$

(d) $Y_Z = Q^{-1}AQX_Z$, where $Q = P^{-1}$.

17. Given the linear transformation $Y_W = \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \\ 1 & 0 & 1 \end{bmatrix} X_W$, relative to the W -basis: $W_1 = [0, -1, 2]'$, $W_2 = [4, 1, 0]'$,

$W_3 = [-2, 0, -4]'$. Find the representation relative to the Z -basis: $Z_1 = [1, -1, 1]'$, $Z_2 = [1, 0, -1]'$, $Z_3 = [1, 2, 1]'$.

$$\text{Ans. } Y_Z = \begin{bmatrix} -1 & 0 & 3 \\ 2 & 2 & -5 \\ -1 & 0 & 2 \end{bmatrix} X_Z$$

18. If, in the linear transformation $Y = AX$, A is singular, then the null space of A is the vector space each of whose vectors transforms into the zero vector. Determine the null space of the transformation of

$$(a) \text{ Problem 12. } (b) \text{ Problem 13. } (c) Y = \begin{bmatrix} 1 & 2 & 3 \\ 2 & 4 & 6 \\ 3 & 6 & 9 \end{bmatrix} X.$$

Ans. (a) $V_3^1(R)$ spanned by $[1, -1, 1]'$
 (b) $V_3^1(R)$ spanned by $[2, 1, -1]'$
 (c) $V_3^2(R)$ spanned by $[2, -1, 0]'$ and $[3, 0, -1]'$

19. If $Y = AX$ carries every vector of a vector space V_n^h into a vector of that same space, V_n^h is called an **invariant space of the transformation**. Show that in the real space $V_3(R)$ under the linear transformation

$$(a) Y = \begin{bmatrix} 1 & 0 & -1 \\ 1 & 2 & 1 \\ 2 & 2 & 3 \end{bmatrix} X, \text{ the } V_3^1 \text{ spanned by } [1, -1, 0]', \text{ the } V_3^1 \text{ spanned by } [2, -1, -2]', \text{ and the } V_3^1 \text{ spanned by } [1, -1, -2]' \text{ are invariant vector spaces.}$$

$$(b) Y = \begin{bmatrix} 2 & 2 & 1 \\ 1 & 3 & 1 \\ 1 & 2 & 2 \end{bmatrix} X, \text{ the } V_3^1 \text{ spanned by } [1, 1, 1]' \text{ and the } V_3^2 \text{ spanned by } [1, 0, -1]' \text{ and } [2, -1, 0]' \text{ are invariant spaces. (Note that every vector of the } V_3^2 \text{ is carried into itself.)}$$

$$(c) Y = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ -1 & 4 & -6 & 4 \end{bmatrix} X, \text{ the } V_4^1 \text{ spanned by } [1, 1, 1, 1]' \text{ is an invariant vector space.}$$

20. Consider the linear transformation $Y = PX$: $y_i = x_{j_i}$, ($i = 1, 2, \dots, n$) in which j_1, j_2, \dots, j_n is a permutation of $1, 2, \dots, n$.

(a) Describe the **permutation matrix** P .

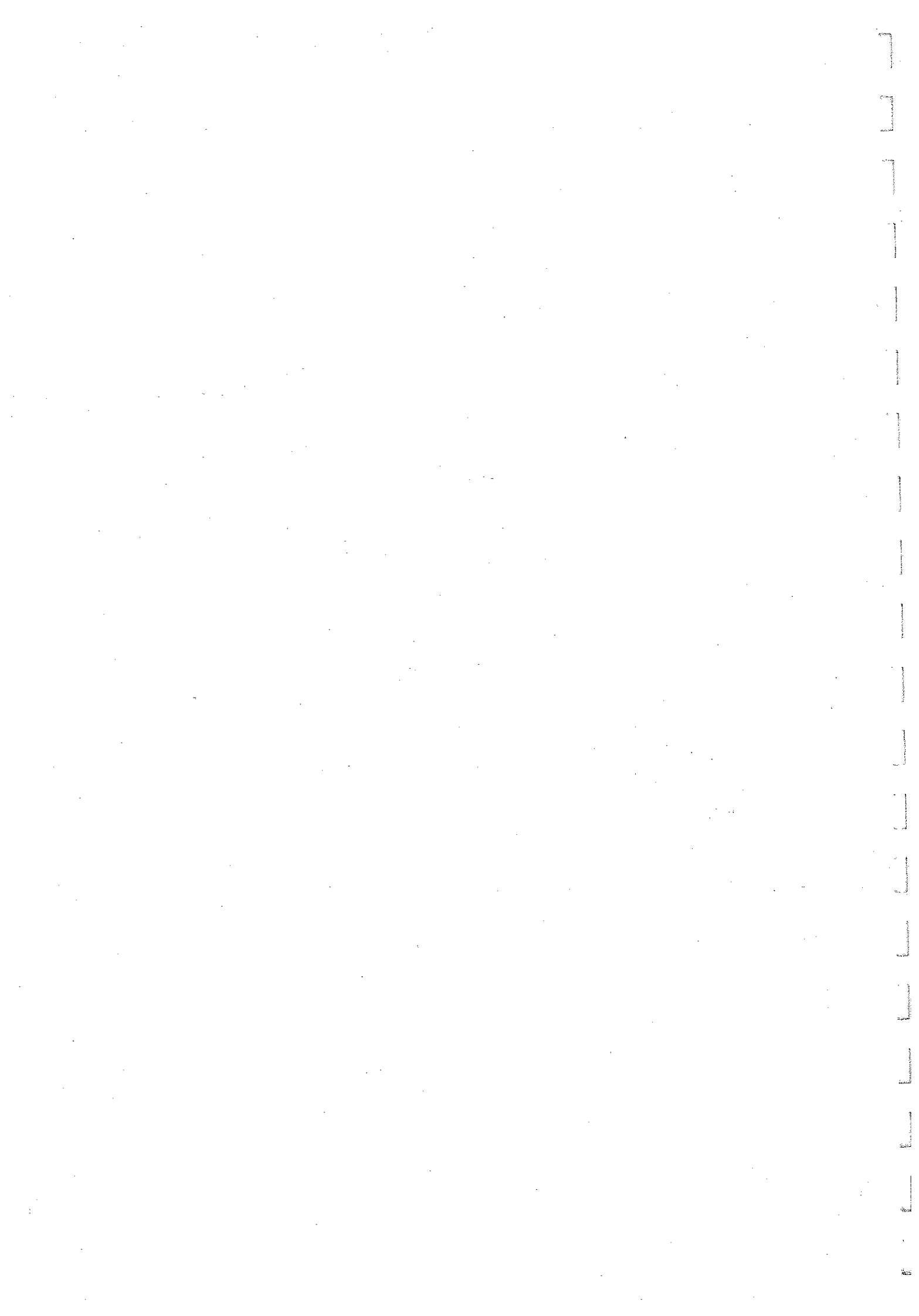
(b) Prove: There are $n!$ permutation matrices of order n .

(c) Prove: If P_1 and P_2 are permutation matrices so also are $P_3 = P_1 P_2$ and $P_4 = P_2 P_1$.

(d) Prove: If P is a permutation matrix so also are P' and $PP' = I$.

(e) Show that each permutation matrix P can be expressed as a product of a number of the elementary column matrices $K_{12}, K_{23}, \dots, K_{n-1, n}$.

(f) Write $P = [E_{i_1}, E_{i_2}, \dots, E_{i_n}]$ where i_1, i_2, \dots, i_n is a permutation of $1, 2, \dots, n$ and E_{i_j} are the elementary n -vectors. Find a rule (other than $P^{-1} = P'$) for writing P^{-1} . For example, when $n = 4$ and $P = [E_3, E_1, E_4, E_2]$, then $P^{-1} = [E_2, E_4, E_1, E_3]$; when $P = [E_4, E_2, E_1, E_3]$, then $P^{-1} = [E_3, E_2, E_4, E_1]$.



INDEX

- Absolute value of a complex number, 110
- Addition
 - of matrices, 2, 4
 - of vectors, 67
- Adjoint of a square matrix
 - definition of, 49
 - determinant of, 49
 - inverse from, 55
 - rank of, 50
- Algebraic complement, 24
- Anti-commutative matrices, 11
- Associative laws for
 - addition of matrices, 2
 - fields, 64
 - multiplication of matrices, 2
- Augmented matrix, 75

- Basis
 - change of, 95
 - of a vector space, 86
 - orthonormal, 102, 111
- Bilinear form(s)
 - canonical form of, 126
 - definition of, 125
 - equivalent, 126
 - factorization of, 128
 - rank of, 125
 - reduction of, 126
- Canonical form
 - classical (Jordan), 206
 - Jacobson, 205
 - of bilinear form, 126
 - of Hermitian form, 146
 - of matrix, 41, 42
 - of quadratic form, 133
 - rational, 203
 - row equivalent, 40
- Canonical set
 - under congruence, 116, 117
 - under equivalence, 43, 189
 - under similarity, 203
- Cayley-Hamilton Theorem, 181
- Chain of vectors, 207
- Characteristic
 - equation, 149
 - polynomial, 149
- Characteristic roots
 - definition of, 149
 - of adj A , 151
 - of a diagonal matrix, 155
 - of a direct sum, 155
- Characteristic roots (cont.)
 - of Hermitian matrices, 164
 - of inverse A , 155
 - of real orthogonal matrices, 155
 - of real skew-symmetric matrices, 170
 - of real symmetric matrices, 163
 - of unitary matrices, 155
- Characteristic vectors
 - (see Invariant vectors)
- Classical canonical form, 206
- Closed, 85
- Coefficient matrix, 75
- Cofactor, 23
- Cogredient transformation, 127
- Column
 - space of a matrix, 93
 - transformation, 39
- Commutative law for
 - addition of matrices, 2
 - fields, 64
 - multiplication of matrices, 3
- Commutative matrices, 11
- Companion matrix, 197
- Complementary minors, 24
- Complex numbers, 12, 110
- Conformable matrices
 - for addition, 2
 - for multiplication, 3
- Congruent matrices, 115
- Conjugate
 - of a complex number, 12
 - of a matrix, 12
 - of a product, 13
 - of a sum, 13
- Conjugate transpose, 13
- Conjunctive matrices, 117
- Contragredient transformation, 127
- Coordinates of a vector, 88
- Cramer's rule, 77

- Decomposition of a matrix into
 - Hermitian and skew-Hermitian parts, 13
 - symmetric and skew-symmetric parts, 12
- Degree
 - of a matrix polynomial, 179
 - of a (scalar) polynomial, 172
- Dependent
 - forms, 69
 - matrices, 73
 - polynomials, 73
 - vectors, 68
- Derogatory matrix, 197

- Determinant
 - definition of, 20
 - derivative of, 33
 - expansion of
 - along first row and column, 33
 - along a row (column), 23
 - by Laplace method, 33
 - multiplication by scalar, 22
 - of conjugate of a matrix, 30
 - of conjugate transpose of a matrix, 30
 - of elementary transformation matrix, 42
 - of non-singular matrix, 39
 - of product of matrices, 33
 - of singular matrix, 39
 - of transpose of a matrix, 21
- Diagonal
 - elements of a square matrix, 1
 - matrix, 10, 156
- Diagonable matrices, 157
- Diagonalization
 - by orthogonal transformation, 163
 - by unitary transformation, 164
- Dimension of a vector space, 86
- Direct sum, 13
- Distributive law for
 - fields, 64
 - matrices, 3
- Divisors of zero, 19
- Dot product, 100

- Eigenvalue, 149
- Eigenvector, 149
- Elementary
 - matrices, 41
 - n -vectors, 88
 - transformations, 39
- Equality of
 - matrices, 2
 - matrix polynomials, 179
 - (scalar) polynomials, 172
- Equations, linear
 - equivalent systems of, 75
 - solution of, 75
 - system of homogeneous, 78
 - system of non-homogeneous, 77
- Equivalence relation, 9
- Equivalent
 - bilinear forms, 126
 - Hermitian forms, 146
 - matrices, 40, 188
 - quadratic forms, 131, 133, 134
 - systems of linear equations, 76

- Factorization into elementary matrices, 43, 188
- Field, 64
- Field of values, 171
- First minor, 22

- Gramian, 103, 111
- Gram-Schmidt process, 102, 111
- Greatest common divisor, 173

- Hermitian form
 - canonical form of, 146
 - definite, 147
 - index of, 147
 - rank of, 146
 - semi-definite, 147
 - signature of, 147
- Hermitian forms
 - equivalence of, 146
- Hermitian matrix, 13, 117, 164
- Hypercompanion matrix, 205

- Idempotent matrix, 11
- Identity matrix, 10
- Image
 - of a vector, 94
 - of a vector space, 95
- Index
 - of an Hermitian form, 147
 - of a real quadratic form, 133
- Inner product, 100, 110
- Intersection space, 87
- Invariant vector(s)
 - definition of, 149
 - of a diagonal matrix, 156
 - of an Hermitian matrix, 164
 - of a normal matrix, 164
 - of a real symmetric matrix, 163
 - of similar matrices, 156
- Inverse of a (an)
 - diagonal matrix, 55
 - direct sum, 55
 - elementary transformation, 39
 - matrix, 11, 55
 - product of matrices, 11
 - symmetric matrix, 58
- Involutory matrix, 11

- Jacobson canonical form, 205
- Jordan (classical) canonical form, 206

- Kronecker's reduction, 136

- Lagrange's reduction, 132
- Lambda matrix, 179
- Laplace's expansion, 33
- Latent roots (vectors), 149
- Leader of a chain, 207
- Leading principal minors, 135
- Left divisor, 180
- Left inverse, 63
- Linear combination of vectors, 68
- Linear dependence (independence)
 - of forms, 70
 - of matrices, 73
 - of vectors, 68
- Lower triangular matrix, 10

- Matrices
 - congruent, 115
 - equal, 2
 - equivalent, 40
 - over a field, 65

- Matrices (cont.)
 product of, 3
 scalar multiple of, 2
 similar, 95, 156
 square, 1
 sum of, 2
- Matrix
 definition of, 1
 derogatory, 197
 diagonal, 157
 diagonal, 10
 elementary row (column), 41
 elementary transformation of, 39
 Hermitian, 13, 117, 164
 idempotent, 11
 inverse of, 11, 55
 lambda, 179
 nilpotent, 11
 non-derogatory, 197
 non-singular, 39
 normal, 164
 normal form of, 41
 nullity of, 87
 of a bilinear form, 125
 of an Hermitian form, 146
 of a quadratic form, 131
 order of, 1
 orthogonal, 103, 163
 periodic, 11
 permutation, 99
 polynomial, 179
 positive definite (semi-definite), 134, 147
 rank of, 39
 scalar, 10
 singular, 39
 skew-Hermitian, 13, 118
 skew-symmetric, 12, 117
 symmetric, 12, 115, 163
 triangular, 10, 157
 unitary, 112, 164
- Matrix polynomial (s)
 definition of, 179
 degree of, 179
 product of, 179
 proper (improper), 179
 scalar, 180
 singular (non-singular), 179
 sum of, 179
- Minimum polynomial, 196
- Multiplication
 in partitioned form, 4
 of matrices, 3
- Negative
 definite form (matrix), 134, 147
 of a matrix, 2
 semi-definite form (matrix), 134, 147
- Nilpotent matrix, 11
- Non-derogatory matrix, 197
- Non-singular matrix, 39
- Normal form of a matrix, 41
- Normal matrix, 164
- Null space, 87
- Nullity, 87
- n -Vector, 85
- Order of a matrix, 1
- Orthogonal
 congruence, 163
 equivalence, 163
 matrix, 103
 similarity, 157, 163
 transformation, 103
 vectors, 100, 110
- Orthonormal basis, 102, 111
- Partitioning of matrices, 4
- Periodic matrix, 11
- Permutation matrix, 99
- Polynomial
 domain, 172
 matrix, 179
 monic, 172
 scalar, 172
 scalar matrix, 180
- Positive definite (semi-definite)
 Hermitian forms, 147
 matrices, 134, 147
 quadratic forms, 134
- Principal minor
 definition of, 134
 leading, 135
- Product of matrices
 adjoint of, 50
 conjugate of, 13
 determinant of, 33
 inverse of, 11
 rank of, 43
 transpose of, 12
- Quadratic form
 canonical form of, 133, 134
 definition of, 131
 factorization of, 138
 rank of, 131
 reduction of
 Kronecker, 136
 Lagrange, 132
 regular, 135
- Quadratic form, real
 definite, 134
 index of, 133
 semi-definite, 134
 signature of, 133
- Quadratic forms
 equivalence of, 131, 133, 134
- Rank
 of adjoint, 50
 of bilinear form, 125
 of Hermitian form, 146
 of matrix, 39
 of product, 43
 of quadratic form, 131
 of sum, 48

- Right divisor, 180
- Right inverse, 63
- Root
 - of polynomial, 178
 - of scalar matrix polynomial, 187
- Row
 - equivalent matrices, 40
 - space of a matrix, 93
 - transformation, 39
- Scalar
 - matrix, 10
 - matrix polynomial, 180
 - multiple of a matrix, 2
 - polynomial, 172
 - product of two vectors (*see* Inner product)
- Schwarz Inequality, 101, 110
- Secular equation (*see* Characteristic equation)
- Signature
 - of Hermitian form, 147
 - of Hermitian matrix, 118
 - of real quadratic form, 133
 - of real symmetric matrix, 116
- Similar matrices, 95, 196
- Similarity invariants, 196
- Singular matrix, 39
- Skew-Hermitian matrix, 13, 118
- Skew-symmetric matrix, 12, 117
- Smith normal form, 188
- Span, 85
- Spectral decomposition, 170
- Spur (*see* Trace)
- Sub-matrix, 24
- Sum of
 - matrices, 2
 - vector spaces, 87
- Sylvester's law
 - of inertia, 133
 - of nullity, 88
- Symmetric matrix
 - characteristic roots of, 163
- Symmetric matrix (cont.)
 - definition of, 12
 - invariant vectors of, 163
- System(s) of equations, 75
- Trace, 1
- Transformation
 - elementary, 39
 - linear, 94
 - orthogonal, 103
 - singular, 95
 - unitary, 112
- Transpose
 - of a matrix, 11
 - of a product, 12
 - of a sum, 11
- Triangular inequality, 101, 110
- Triangular matrix, 10, 157
- Unit vector, 101
- Unitary
 - matrix, 112
 - similarity, 157
 - transformation, 112
- Upper triangular matrix, 10
- Vector(s)
 - belonging to a polynomial, 207
 - coordinates of, 88
 - definition of, 67
 - inner product of, 100
 - invariant, 149
 - length of, 100, 110
 - normalized, 102
 - orthogonal, 100
 - vector product of, 109
- Vector space
 - basis of, 86
 - definition of, 85
 - dimension of, 86
 - over the complex field, 110
 - over the real field, 100

Index of Symbols

Symbol	Page	Symbol	Page
a_{ij}	1	E_i , (vector)	88
$[a_{ij}]$	1	$X \cdot Y$; $X Y$	100, 110
A	1	$\ X\ $	100, 110
Σ	3	G	103, 111
I, I_n	10	$X \times Y$	109
A^{-1} ; A^I	11	\mathcal{L}	115
A' ; A^T	11	p	116
\bar{A} ; A^C	12	s	116
\bar{A}' ; A^* ; A^{CT}	13	q	131
$ A $; $\det A$	20	h	146
$ M_{ij} $	22	λ, λ_i	149
$A_{i_1, i_2, \dots, i_m}^{j_1, j_2, \dots, j_m}$	23	$\phi(\lambda)$	149
α_{ij}	23	E_i , (matrix)	170
r	39	$f(\lambda)$	172
H_{ij}, K_{ij}	39	$F[\lambda]$	172
$H_i(k), K_i(k)$	39	$A(\lambda)$	179
$H_{ij}(k), K_{ij}(k)$	39	$A_R(C), A_L(C)$	180
\sim	40	$N(\lambda)$	189
N	43	$f_i(\lambda)$	189
$\text{adj } A$	49	$m(\lambda)$	196
F	64	$C(g)$	198
X, X_i	67	J	198
$V_n(F)$	85	S	203
$V_n^m(F)$	86	$C_q(p)$	205
N_A	87		

