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ADI BEN-ISRAEL

**On Decompositions of Matrix Spaces with  
Applications to Matrix Equations**

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**Algebra lineare.** — *On Decompositions of Matrix Spaces with Applications to Matrix Equations*<sup>(\*)</sup>. Nota <sup>(\*\*)</sup> di ADI BEN-ISRAEL, presentata dal Socio B. SEGRE.

RIASSUNTO. — Si dimostra un teorema di decomposizione dello spazio  $C^{m \times n}$ , delle matrici complesse  $m \times n$ , in una somma diretta di sottospazi ortogonali complementari, come conseguenza di un teorema corrispondente dello spazio vettoriale di dimensione  $mn$ . Si danno inoltre delle applicazioni per equazioni matriciali.

#### INTRODUCTION.

A theorem on the decomposition of  $C^{m \times n}$ , the space of  $m \times n$  complex matrices, into a direct sum of orthogonal complementary subspaces [1] is proved here as a consequence of the corresponding theorem in  $C^{mn}$ , the  $mn$  dimensional complex vector space. Applications to matrix equations are given.

#### § 0. — NOTATIONS.

$C^n$  the  $n$ -dimensional complex vector space

$(x, y) = \sum_{i=1}^n x_i \bar{y}_i$  the standard inner product in  $C^n$

$\|x\|_2 = (x, x)^{1/2}$  the Euclidean norm in  $C^n$ .

For any subspace  $L$  of  $C^n$ :

$L^\perp$  the orthogonal complement of  $L$

$C^n = L \oplus M$  denotes  $M = L^\perp$

$C^{m \times n}$  the space of  $m \times n$  complex matrices.

For any  $A \in C^{m \times n}$ :

$A^t$  the transpose of  $A$

$A^*$  the conjugate transpose of  $A$

$A^+$  the generalized inverse of  $A$ . [6]

$R(A)$  the range of  $A$

$N(A)$  the null space of  $A$ .

For any subspace  $L$  of  $C^n$ :

$P_L$  the perpendicular projection on  $L$

i.e.  $P_L = P_L^2 = P_L^*$ ,  $R(P_L) = L$ .

For any  $A \in C^{m \times n}$ ,  $B \in C^{p \times q}$  the Kronecker product of  $A$ ,  $B$  is

$A \otimes B = (a_{ij} B) \in C^{mp \times nq}$  ( $i = 1, \dots, m$ ;  $j = 1, \dots, n$ ), [5].

If not specified, the dimensions of matrices should be clear from the context.

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§ 1. — A CORRESPONDENCE BETWEEN  $C^{m \times n}$  AND  $C^{mn}$ .

Aside from the practical aspect of representing matrices on tapes or punched cards, there seems to be little interest or use in the observation that any  $m \times n$  matrix may be regarded as an  $mn$ -dimensional vector. While most of the interesting matrix properties are lost in passing from  $C^{m \times n}$  to  $C^{mn}$ , those vector properties of linearity, convexity, standard inner product and the Euclidean norm are naturally preserved by the following correspondence.

*Definition 1:* Let  $v: C^{m \times n} \rightarrow C^{mn}$  be the mapping assigning to any

$$X = (x_{ij}) \in C^{m \times n} \quad \text{the vector} \quad v(X) = (v_k), \quad (k = 1, \dots, mn),$$

$$\text{given by} \quad v_{n(i-1)+j} = x_{ij} \quad (i = 1, \dots, m; j = 1, \dots, n)$$

i.e.  $v(X)$  is the vector obtained by reading the rows of  $X$  one by one.

The mapping  $v$  induces in  $C^{m \times n}$  the inner product

$$(1) \quad (X, Y) = (v(X), v(Y)) = \sum_{i,j} x_{ij} \bar{y}_{ij} = \text{trace } Y^* X$$

and the norm

$$(2) \quad \|X\| = \|v(X)\|_2 = \left( \sum_{i,j} \|x_{ij}\|^2 \right)^{1/2}.$$

Since  $v: C^{m \times n} \rightarrow C^{mn}$  is a nonsingular linear transformation it is clear that  $L$  is a subspace of  $C^{m \times n}$  if and only if  $v(L)$  is a subspace of  $C^{mn}$ , and that  $\dim L = \dim v(L)$ . The following subspaces in  $C^{m \times n}$  are of special interest:

*Definition 2:* For any  $A \in C^{m \times p}$ ,  $B \in C^{q \times n}$

$$(3) \quad R(A, B) = \{X : X = AYB \text{ for some } Y \in C^{p \times q}\}$$

the *range* of  $(A, B)$

$$(4) \quad N(A, B) = \{Y : AYB = 0\}$$

the *null space* of  $(A, B)$ .

The vector space counterparts of these subspaces are given in:

LEMMA 1:

$$(i) \quad v(R(A, B)) = R(A \otimes B')$$

$$(ii) \quad v(N(A, B)) = N(A \otimes B').$$

*Proof.*—Follows from the easily verified

$$(5) \quad v(AYB) = (A \otimes B') v(Y), \quad \text{for all } Y, \quad \text{e.g. [5] p. 9.}$$

Not all the subspaces in  $C^{m \times n}$  are of the form (3) or (4) since not all the  $mn \times pq$  matrices are of the form  $A \otimes B'$ ,  $A \in C^{m \times p}$ ,  $B \in C^{q \times n}$ . For example, the subspace of symmetric matrices in  $R^{n \times n}$  can be represented in the form (3) but only after rearrangement of components.

§ 2. — ON RANGE-NULL SPACE DECOMPOSITIONS OF  $C^{m \times n}$ .

For any  $A \in C^{m \times n}$  we recall that

$$(6) \quad C^m = R(A) \oplus N(A^*)$$

The analogous result in  $C^{m \times n}$  is:

THEOREM 1 ([1], [3]).—For any  $A \in C^{m \times p}$ ,  $B \in C^{q \times m}$

$$(7) \quad C^{m \times n} = R(A, B) \oplus N(A^*, B^*).$$

*Proof.*—Follows from lemma 1 since, by (6), the subspaces

$$R(A \otimes B^t) = v(R(A, B))$$

and

$$N((A \otimes B^t)^*) = N(A^* \otimes B^{*t}) = v(N(A^*, B^*))$$

are orthogonal complements in  $C^{mn}$ .

This theorem may be stated more generally [1], but the restriction to matrices makes possible the above elementary derivation.

Before giving the perpendicular projections corresponding to the decomposition (7) we need:

LEMMA 2.—Let  $S, T, S_i, T_i$  ( $i = 1, \dots, k$ ) be matrix spaces and let

$$f: \prod_{i=1}^k S_i \rightarrow S, : \prod_{i=1}^k T_i \rightarrow T$$

be a mapping satisfying:

$$(i) \quad \text{For all } A_i \in S_i, B_i \in T_i \ (i = 1, \dots, k)$$

$$f(A_1, \dots, A_k) f(B_1, \dots, B_k) = f(C_1, \dots, C_k)$$

where for  $i = 1, \dots, k$

$$C_i = A_i B_i \quad \text{or} \quad B_i A_i \quad (1)$$

$$(ii) \quad \text{If } A_i \ (i = 1, \dots, k) \text{ are Hermitian then so is } f(A_1, \dots, A_k).$$

Then:

$$(8) \quad (f(A_1, \dots, A_k))^+ = f(A_1^+, \dots, A_k^+)$$

for all

$$A_i \in S, \quad (i = 1, \dots, k).$$

*Proof.*—The right side of (8) satisfies the defining conditions of the generalized inverse of  $f(A_1, \dots, A_k)$ , e.g. [6].

(1) One choice for each  $i$  but possibly different choices for different  $i$ , e.g.  $f(A_1, A_2) f(B_1, B_2) = f(A_1 B_1, B_2 A_2)$ .

COROLLARY 1.—For any matrices  $A, B$ .

$$(i) \quad (A^t)^+ = (A^+)^t \quad (2)$$

$$(ii) \quad (A \otimes B)^+ = A^+ \otimes B^+$$

*Proof.*—Use lemma 2 with:

$$(i) \quad k = 1, f(A) = A^t$$

$$(ii) \quad k = 2, f(A, B) = A \otimes B$$

and verify in each case that  $f$  satisfies conditions (i), (ii) of lemma 2.

COROLLARY 2.—The perpendicular projections of  $C^{m \times n}$  on the subspaces  $R(A, B), N(A^*, B^*)$  of theorem 1 are given by:

$$(9) \quad P_{R(A, B)} X = AA^+ XB^+ B$$

$$(10) \quad P_{N(A^*, B^*)} X = X - AA^+ XB^+ B$$

for any  $X \in C^{m \times n}$

*Proof.*—(9) follows from (5) and lemma 1 since

$$\begin{aligned} P_{R(A \otimes B^t)} &= (A \otimes B^t) (A \otimes B^t)^+ && \text{e.g. [3]} \\ &= (A \otimes B^t) (A^+ \otimes B^{+t}) && \text{by corollary 1} \\ &= (AA^+) \otimes (B^+ B)^t && \text{e.g. [5]} \end{aligned}$$

(10) follows now from (9) and (7).

The projection (9) is rewritten as

$$P_{R(A, B)} X = P_{R(A)} X P_{R(B^*)}$$

and (10), by subtracting and adding  $AA^+ X$ , becomes

$$P_{N(A^*, B^*)} X = P_{N(A^*)} X + P_{R(A)} X P_{N(B)}$$

or alternatively

$$P_{N(A^*, B^*)} X = P_{N(A^*)} X P_{R(B^*)} + X P_{N(B)}.$$

The corresponding projections in  $C^{mn}$  are therefore

$$\begin{aligned} P_{v(R(A, B))} &= P_{R(A \otimes B^t)} = P_{R(A)} \otimes P_{R(B^*)}^t \\ P_{v(N(A^*, B^*))} &= P_{N(A^* \otimes B^{*t})} = P_{N(A^*)} \otimes I + P_{R(A)} \otimes P_{N(B)}^t \\ &= P_{N(A^*)} \otimes P_{R(B^*)}^t + I \otimes P_{N(B)}^t. \end{aligned}$$

(2) This is different from

$$(A^*)^+ = (A^+)^*, [6]$$

which also can be proved by lemma 2.

## § 3. — APPLICATIONS.

The above results have direct applications to matrix equations:

THEOREM 2 (PENROSE [6]).—*The matrix equation*

$$(11) \quad AXB = C$$

*is solvable if, and only if*

$$(12) \quad AA^+CB^+B = C$$

*in which case the general solution is*

$$(13) \quad A^+CB^+ + Y - A^+AYBB^+, \quad Y \text{ arbitrary}$$

*Proof.*—(11) is solvable if, and only if  $C \in R(A, B)$  which proves (12) by using (9). The general solution is any particular solution, e.g.  $A^+CB^+$  by (12), plus the general element of  $N(A, B)$  which by (10) proves (13). The least squares solution of (11) are also easily obtainable from the above results:

THEOREM 3 (PENROSE [7]).—*The matrix*

$$(14) \quad A^+CB^+$$

*is of minimal norm (2) among all matrices minimizing*

$$\|AXB - C\|.$$

*Proof.*—It follows from the corresponding result in  $C^{mn}$  that the vector  $(A \otimes B^t)^+ v(C) = (A^+ \otimes B^+)^t v(C) = v(A^+CB^+)$  is of minimal norm among all vectors minimizing

$$\|(A \otimes B^t) v(X) - v(C)\| = \|v(AXB - C)\|.$$

The following characterization of  $A^+$  is also interesting:

COROLLARY 3.—*Let  $A \in C^{m \times n}$  and  $X$  satisfy*

$$(15) \quad AXA = A.$$

*Then the following are equivalent:*

- (i)  $X = A^+$
- (ii)  $X \in R(A^*, A^*)$
- (iii)  $X$  is the minimal norm (2) solution of (15).

*Proof.*—The general solution of (15) is

$$(16) \quad \begin{aligned} X &= A^+AA^+ + Y - A^+AYAA^+, & \text{by (13)} \\ &= A^+ + P_{N(A, A)} Y, & \text{by (10)} \end{aligned}$$

(i)  $\iff$  (ii) now follows from (7) since

$$A^+ = A^+AA^+AA^+ = A^*A^{*+}A^+A^{*+}A^* \in R(A^*, A^*)$$

and (i)  $\iff$  (iii) from

$$\|X\|^2 = \|A^+\|^2 + \|P_{N(A,A)} Y\|^2 \quad \text{in (16).}$$

An application to matrix inequalities will now be given. For any  $X = (x_{ij}) \in \mathbb{R}^{m \times n}$  we denote by  $X \geq 0$  the fact

$$x_{ij} \geq 0 \quad (i = 1, \dots, m; j = 1, \dots, n).$$

**COROLLARY 4.**—*Let  $A, B, C$  be real matrices. Then the system of equations and inequalities*

$$(17) \quad AXB = C, \quad X \geq 0$$

*is solvable if, and only if,*

$$(18) \quad A^t U B^t \geq 0 \quad \text{implies: } \text{trace } U^t C \geq 0.$$

*Proof.*—The solvability of (17) is equivalent to that of

$$(A \otimes B^t) v(X) = v(C), \quad v(X) \geq 0$$

which by Farkas' theorem [4] is equivalent to:

$$(A \otimes B^t)^t v(U) \geq 0 \quad \text{implies } (v(U), v(C)) \geq 0$$

or, by (5) and (1), to (18).

Applications of these results to iterative methods of generalized inversion are given in [10]. In particular it is shown that for  $X_0 \in \mathbb{R}(A^*, A^*)$  the iterative method [2] (or the higher order methods of [8], [9]):

$$X_{k+1} = X_k (2I - AX_k) \quad (k = 0, 1, \dots)$$

converges to  $A^+$  if, and only if the spectral radius:

$$\rho(P_{R(A)} - AX_0) < 1,$$

but that it may diverge for  $X_0$  with  $P_{N(A,A)} X_0 \neq 0$  even if

$$\rho(P_{R(A)} - AX_0) = 0$$

$$\text{e.g. } A = \frac{1}{2} \begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix}, \quad X_0 = \frac{1}{2} \begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix} + \epsilon \begin{pmatrix} 1 & -1 \\ -1 & 1 \end{pmatrix}, \quad \epsilon \neq 0.$$

#### REFERENCES.

- [1] A. BEN-ISRAEL, *On direct sum decompositions of Hestenes algebras*, « Israel J. Math. », 2, 50-54 (1964).
- [2] A. BEN-ISRAEL, *A note on an iterative method for generalized inversion of matrices*, « Math. Comp. », 20, 439-440 (1966).
- [3] A. BEN-ISRAEL and A. CHARNES, *Contributions to the theory of generalized inverses*, « J. SIAM », 11, 667-699 (1963).
- [4] A. CHARNES and W. W. COOPER, *Management models and industrial applications of linear programming*, vols. I-II, J. Wiley, New York 1961.

- [5] M. MARCUS and H. MINC, *A survey of matrix theory and matrix inequalities*, Allyn & Bacon, xvi + 180 pp., Boston 1964.
- [6] R. PENROSE, *A generalized inverse for matrices*, « Proc. Cambridge Philos. Soc. », 51, 406–413 (1955).
- [7] R. PENROSE, *On best approximate solutions of linear matrix equations*, « Proc. Cambridge Philos. Soc. », 52, 17–19 (1956).
- [8] W. V. PETRYSHYN, *On generalized inverses and on the uniform convergence of  $(I - \beta K)^n$  with applications to iterative methods*, « J. Math. Anal. Appl. », 18, 417–439 (1967).
- [9] S. ZLOBEC, *On computing the generalized inverse of a linear operator*, « Glasnik Mat. », 22, 265–271 (1967).
- [10] S. ZLOBEC and A. BEN-ISRAEL, *On  $K_u$ -symmetric and  $K_u$ -p.d. matrices and the iterative computation of generalized inverses*, Systems Research Memorandum No 214, Northwestern University, Evanston, Ill., July 1968.