

## SUM OF SECONDARY RANGE SYMMETRIC MATRICES

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ABSTRACT. A necessary and sufficient condition for sum of secondary range symmetric matrices to be secondary range symmetric is given here. Also, the concept of parallel summable secondary range symmetric matrices and some of its characterizations are being explored.

### 1. Introduction

The concepts of secondary symmetric matrices is introduced by [4]. Vijayakumar et.al [2], [3], [12] defined the concept of secondary normal matrices and its characterizations. Based on this concept, secondary generalized inverses and its characterizations are obtained in [7]. Drazin theta inverse [10] is a generalized inverse of a matrix  $A$ , obtained by combining Drazin inverse as well as the secondary conjugate transpose of the matrix  $A$ . For more properties and definitions involving secondary (conjugate) transpose, one can refer [10], [8], [9].

In this article, we give essential conditions for sum of secondary range symmetric matrices to be secondary symmetric. Also, the parallel summability of secondary range symmetric matrices are obtained. Below are some preliminary results from the literature.

### 2. Preliminaries

Throughout this article,  $\mathbb{C}^{m \times n}$  represents the set of all matrices over complex field.

**Definition 2.1.** [1]  $A^-$  is a generalized inverse (g-inverse) of  $A$  if  $AA^-A = A$ .

**Definition 2.2.** [1] A matrix  $A \in \mathbb{C}^{n \times n}$  is range symmetric (or EP) if  $Null(A) = Null(A^*)$ .

**Definition 2.3.** [7]  $A^{\dagger_s}$  is a secondary generalized inverse of  $A$  if  $AA^{\dagger_s}A = A$ ,  $A^{\dagger_s}AA^{\dagger_s} = A$  and  $AA^{\dagger_s}$ ,  $A^{\dagger_s}A$  are secondary symmetric.

**Definition 2.4.** [2] Let  $A \in \mathbb{C}^{n \times n}$ . Then the conjugate secondary transpose of  $A$  denoted by  $A^\theta$  and is defined as  $A^\theta = \overline{A}^s = (c_{ij})$  where  $c_{ij} = \overline{a_{n-j+1, n-i+1}}$ .

Whenever we consider the matrix on set of real numbers, the secondary conjugate transpose  $A^\theta$  reduces to secondary transpose  $A^S$ .

A relation connecting the transpose and secondary transpose of the matrix  $A$  is,

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$A^S = VA^TV$  where  $V$  is a permutation matrix having unity in the secondary diagonal.

**Definition 2.5.** [11] A matrix  $A \in \mathbb{R}^{n \times n}$  secondary range symmetric if and only if  $\text{Null}(A) = \text{Null}(A^S)$

**Theorem 2.6.** [11] Let  $A \in \mathbb{R}^{n \times n}$ . Then the following conditions are equivalent.

- (i)  $A$  is secondary range symmetric
- (ii)  $VA$  is range symmetric
- (iii)  $AV$  is range symmetric
- (iv)  $\text{Null}(A^*) = \text{Null}(AV)$
- (v)  $\mathcal{C}(A) = \mathcal{C}(A^S)$
- (vi)  $A^S = PA = AQ$  where  $P$  and  $Q$  are some nonsingular matrices
- (vii)  $\mathcal{C}(A^*) = \mathcal{C}(VA)$
- (viii)  $\mathcal{C}(A^*) \oplus \text{Null}(A) = \mathbb{C}^n$
- (ix)  $\mathcal{C}(A) \oplus \text{Null}(A^*) = \mathbb{C}^n$

### 3. Results

**Lemma 3.1.** Let  $A_1, A_2, \dots, A_k \in \mathbb{R}^{n \times n}$ . If  $A = \sum_{i=1}^k A_i$ , then  $A^S = \sum_{i=1}^k A_i^S$

*Proof.* By definition  $A_i^S = VA_i^*V$  for  $i = 1, 2, \dots, k$ .

To prove  $A^S = \sum_{i=1}^k A_i^S$ ,

Given  $A = \sum_{i=1}^k A_i$ .

Now,

$$\begin{aligned} A^S &= V(A_1 + A_2 + \dots + A_k)^*V \\ &= V(A_1^* + A_2^* + \dots + A_k^*)V \\ &= A_1^S + A_2^S + \dots + A_k^S \end{aligned}$$

Hence  $A^S = \sum_{i=1}^k A_i^S$ . □

**Lemma 3.2.** Let  $A, B \in \mathbb{R}^{n \times n}$ , then

- (i)  $(AB)^S = B^S A^S$
- (ii)  $(A^S)^S = A$

*Proof.*

$$\begin{aligned} \text{By definition } (AB)^S &= V(AB)^*V \\ &= V(B^*A^*)V \\ &= (VB^*V)(VA^*V) \\ &= B^S A^S \end{aligned}$$

(ii) follows from (i) since  $(A^S)^S = (VA^*V)^S = V(A^*)^S V = A$ . □

**Lemma 3.3.** [5] Let  $A_1, A_2, \dots, A_k \in \mathbb{C}^{n \times n}$  and let  $A = \sum_{i=1}^k A_i$ . Consider the following conditions.

- (i)  $\text{Null}(A) \subseteq \text{Null}(A_i)$ ;  $i = 1, 2, \dots, k$

$$(ii) \text{ Null}(A) = \bigcap_{i=1}^k \text{Null}(A_i)$$

$$(iii) \text{ rank}(A) = \text{rank} \begin{pmatrix} A_1 \\ A_2 \\ \vdots \\ A_k \end{pmatrix}$$

$$(iv) \sum_{i=1}^k \sum_{j=1}^n A_i^* A_j = 0$$

$$(v) \text{rank}(A) = \sum_{i=1}^k \text{rank}(A_i)$$

Then the following statement hold:

(a) (i), (ii), (iii) are equivalent.

(b) conditions (iv) implies (i) but the converse is not true.

(c) conditions (v) implies (i) but not the converse.

**Theorem 3.4.** Let  $A_i (i = 1, 2, \dots, k)$  be secondary range symmetric. Assume any one of the condition of previous lemma holds. Then  $A = \sum_{i=1}^k A_i$  is secondary range symmetric.

*Proof.* Since each  $A_i$  is secondary range symmetric, by definition 2.5,  $\text{Null}(A_i) = \text{Null}(A_i^S)$  for each  $i = 1, 2, \dots, k$ . By the given condition

$$\text{Null}(A) \subseteq \text{Null}(A_i)$$

we get

$$\text{Null}(A) \subseteq \bigcap_{i=1}^k \text{Null}(A_i) = \bigcap_{i=1}^k \text{Null}(A_i^S)$$

Now,

$$\begin{aligned} x \in \text{Null}(A) \subseteq \bigcap_{i=1}^k \text{Null}(A_i^S) &\implies x \in \text{Null}(A_i^S), \text{ for } i = 1 \text{ to } k \\ &\implies A_i^S x = 0, \text{ for } i = 1 \text{ to } k \\ &\implies (A_1^S + A_2^S + \dots + A_k^S)x = 0 \\ &\implies A^S x = 0 \end{aligned}$$

$$\bigcap_{i=1}^k \text{Null}(A_i^S) \subseteq \text{Null}(A^S)$$

$$\text{Null}(A) \subseteq \bigcap_{i=1}^k \text{Null}(A_i^S) \subseteq \text{Null}(A^S) \text{ and } \rho(A) = \rho(A^S) \text{ implies } \text{Null}(A) =$$

$$\text{Null}(A^S). \text{ Thus } A = \bigcap_{i=1}^k A_i \text{ is secondary range symmetric. } \quad \square$$

The converse of the above theorem is not true.

$$\text{Let } A_1 = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \text{ and } A_2 = \begin{pmatrix} 0 & -1 \\ 0 & 0 \end{pmatrix}.$$

$$\text{Assume } A = A_1 + A_2 = \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix}.$$

Here,  $A_1, A_2$  and  $A$  are secondary range symmetric matrices.

But  $Null(A_1 + A_2) \not\subseteq Null(A)$ .

If  $A$  and  $B$  are secondary range symmetric matrices, then by Theorem 2.6  $A^S = P_1A$  and  $B^S = P_2A$  where  $P_1$  and  $P_2$  are non singular square matrices. If  $P_1 = P_2$ , say  $P$ , then  $A^S + B^S = P(A + B) = (A + B)^S$  which implies  $A + B$  is secondary range symmetric.

**Theorem 3.5.** *Let  $A$  and  $B$  be secondary range symmetric matrices, then  $A^S = P_1A$  and  $B^S = P_2B$  such that  $P_1 - P_2$  is a non singular matrix. Then  $A + B$  is secondary range symmetric iff  $Null(A + B) \subseteq Null(B)$ .*

*Proof.* Since  $A^S = P_1A$  and  $B^S = P_2B$ , from theorem 2.6, the matrices  $A$  and  $B$  are secondary range symmetric. Since  $Null(A + B) \subseteq Null(B)$ , we can see that  $Null(A + B) \subseteq Null(A)$ . Hence by theorem 3.4,  $A + B$  is secondary range symmetric.

Conversly, let us assume that  $A + B$  is secondary range symmetric. Now by Theorem 2.6,  $(A + B)^S = P(A + B)$  which implies

$$P(A + B) = A^S + B^S = P_1A + P_2B$$

$$(P_1 - P)A = (P - P_2)B$$

$$EA = FB$$

where  $E = P_1 - P$  and  $F = P - P_2$  such that  $E + F = P_1 - P_2$ .

$$EA + FA = EB + FA$$

$$(E + F)A = F(A + B)$$

By hypothesis  $E + F = H_1 - H_2$  is non singular.  $Null(A + B) \subseteq Null[F(A + B)] = Null[(E + F)A] = Null(A)$ . Similarly we can see that  $Null(A + B) \subseteq Null(B)$ . Thus  $A + B$  is secondary range symmetric implies  $Null(A + B) \subseteq Null(A)$  and  $Null(B)$ . Hence the Theorem.  $\square$

#### 4. Parallel summable secondary range symmetric matrices

**Definition 4.1.** [6] Let  $A, B \in \mathbb{C}^{m \times n}$  matrices. If  $Null(A + B) \subseteq Null(B)$  and  $Null(A + B)^* \subseteq Null(B^*)$  or  $Null(A + B) \subseteq Null(A)$  and  $Null(A + B)^* \subseteq Null(A^*)$  then the matrices  $A$  and  $B$  are said to be parallel summable.

**Definition 4.2.** [6] If two matrices  $A$  and  $B$  are parallel summable, then the parallel sum of is denoted by  $A : B$  is defined as  $A : B = A(A + B)^-B$ .

**Lemma 4.3.** *Let  $A$  and  $B$  be matrices. Then  $Null(A^*) \subseteq Null(B^*)$  if and only if  $Null(A^S) \subseteq Null(B^S)$*

*Proof.* Assume that  $Null(A^*) \subseteq Null(B^*)$

$$\begin{aligned}
 x \in Null(A^S) &\implies A^S x = 0 \\
 &\implies VA^*Vx = 0 \\
 &\implies A^*Vx = 0 \\
 &\implies A^*y = 0 \text{ where } y = Vx \\
 &\implies y \in Null(A^*) \subseteq Null(B^*) \\
 &\implies B^*y = 0 \\
 &\implies B^*Vx = 0 \\
 &\implies B^S x = 0 \\
 &\implies x \in Null(B^S)
 \end{aligned}$$

Thus  $Null(A^S) \subseteq Null(B^S)$ .

Conversly, let us assume that  $Null(A^S) \subseteq Null(B^S)$ . We need to show that  $Null(A^*) \subseteq Null(B^*)$ .

Let us choose

$$\begin{aligned}
 x \in Null(A^*) &\implies A^*x = 0 \\
 &\implies (VA^*V)Vx = 0 \\
 &\implies A^S Vx = 0 \\
 &\implies A^S y = 0 \\
 &\implies y \in Null(A^S) \subseteq Null(B^S) \\
 &\implies B^S y = 0 \\
 &\implies VB^*VVx = 0 \\
 &\implies VB^*x = 0 \\
 &\implies B^*x = 0 \\
 &\implies x \in Null(B^*)
 \end{aligned}$$

Thus  $Null(A^*) \subseteq Null(B^*)$ . Hence the Lemma.  $\square$

**Definition 4.4.** A pair of matrices  $A$  and  $B$  are said to be secondary parallel summable if  $Null(A+B) \subseteq Null(B)$  and  $Null(A+B)^S \subseteq Null(B^S)$  or equivalently  $Null(A+B) \subseteq Null(A)$  and  $Null(A+B)^S \subseteq Null(A^S)$

## 5. Properties

Let  $A$  and  $B$  be a pair of secondary parallel summable matrices.

- (i)  $A : B$
- (ii)  $A^S$  and  $B^S$  are secondary parallel summable and  $(A : B)^S = A^S : B^S$
- (iii) If  $C$  is a non singular matrix, then  $CA$  and  $CB$  are parallel summable and  $CA : CB = C(A : B)$
- (iv)  $Null(A : B) = Null(A) + Null(B)$

*Proof.* Here we need to prove only (ii),  $A$  and  $B$  are secondary range symmetric,  $A^S$  and  $B^S$  are range symmetric follows from Lemma 3.1 and 3.2.

$$\begin{aligned}
 A^S : B^S &= A^S(A^S + B^S)^- B^S \\
 &= A^S[(A + B)^S]^- B^S \\
 &= A^S[(A + B)^-]^S B^S \\
 &= [B(A + B)^- A]^S \\
 &= [A(A + B)^- B]^S \\
 &= [A : B]^S
 \end{aligned}$$

□

**Lemma 5.1.** *Let  $A$  and  $B$  be secondary range symmetric matrices. Then  $A$  and  $B$  are parallel summable secondary range symmetric if and only if  $\text{Null}(A + B) \subseteq \text{Null}(A)$*

*Proof.*  $A$  and  $B$  are parallel summable, by definition 4.1 it follows that  $\text{Null}(A + B) \subseteq \text{Null}(A)$ .

Conversly, if  $\text{Null}(A + B) \subseteq \text{Null}(A)$ , then  $\text{Null}(A + B) \subseteq \text{Null}(B)$ . Since  $A$  and  $B$  are secondary range symmetric,  $A + B$  is secondary range symmetric by Theorem 3.4. Hence  $\text{Null}(A + B) = \text{Null}(A + B)^S$  and  $\text{Null}(A + B) \subseteq \text{Null}(A)$  implies  $\text{Null}(A + B)^S \subseteq \text{Null}(A^S)$ . Then by Definition 4.1,  $A$  and  $B$  are parallel summable secondary range symmetric. Hence the proof. □

If  $A$  or  $B$  is not secondary range symmetric matrix, the above lemma is not true. Consider the following example:

Let  $A = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}$  and  $A^S = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}$ . Here  $A$  is secondary range symmetric matrix.

Also, let  $B = \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix}$  and  $B^S = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}$ . It is clear that  $B$  is not secondary range symmetric.

Now,  $A + B = \begin{pmatrix} 0 & 1 \\ 0 & 1 \end{pmatrix}$  and  $(A + B)^S = \begin{pmatrix} 1 & 1 \\ 0 & 0 \end{pmatrix}$ . Here  $\text{Null}(A + B) \subseteq \text{Null}(A)$ , but  $\text{Null}(A + B)^S \not\subseteq \text{Null}(A^S)$  and  $\text{Null}(B^S)$ . Hence  $A$  and  $B$  are not secondary parallel summable.

**Theorem 5.2.** *Let  $A$  and  $B$  be parallel summable secondary range symmetric. Then  $A : B$  and  $A + B$  are secondary range symmetric.*

*Proof.* Since  $A$  and  $B$  are parallel summable secondary range symmetric,  $\text{Null}(A + B) \subseteq \text{Null}(A)$  and  $\text{Null}(A + B) \subseteq \text{Null}(B)$ , follows from above lemma. Now the fact that  $A + B$  is secondary range symmetric follows from Theorem 3.4.  $A : B$  is secondary range symmetric runs as follows.

$$\begin{aligned}
Null(A : B)^S &= Null(A^S : B^S) \\
&= Null(A^S) + Null(B^S) \\
&= Null(A) + Null(B) \\
&= Null(A : B)
\end{aligned}$$

Thus  $A : B$  is secondary range symmetric whenever  $A$  and  $B$  are parallel summable secondary range symmetric. Hence the theorem.  $\square$

## 6. Conclusion

In general, the sum of two secondary range symmetric matrices need not be secondary range symmetric. We obtained a necessary and sufficient condition for sum of secondary range symmetric matrices to be secondary range symmetric. Similarly, the essential conditions for product of secondary range symmetric matrices to be secondary range symmetric can be a topic for further research.

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