

## Secondary k-Generalized Inverse of a s-k-Normal Matrices

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### Abstract:

Secondary k-generalized inverse of a given square matrix is defined and its characterizations are given. Secondary k-generalized inverses of s-k normal matrices are discussed.

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### 1.Introduction:

Ann Lee initiated the study of secondary symmetric matrices in [1]. The concept of secondary k-normal matrices was introduced in [3]. Some equivalent conditions on secondary k-normal matrices are given in [4]. In this paper we describe secondary k-generalized inverse of a square matrix, as the unique solution of a certain set of equation. This secondary k-generalized inverse exists for particular kind of square matrices. Let  $C_{n \times n}$  denote the space of  $n \times n$  complex matrices. We deal with secondary k-generalized inverse of s-k normal matrices. Through this paper, if  $A \in C_{n \times n}$ , then we

assume that if  $A \neq 0$  then  $A(KVA^*VK) \neq 0$

$$\text{i.e., } A(KVA^*VK) = 0 \Rightarrow A = 0 \rightarrow (1)$$

It is clear that the conjugate secondary k transpose satisfies the following properties.

$$KV(A+B)^*VK = (KVA^*VK) + (KVB^*VK)$$

$$KV(\lambda A)^*VK = \bar{\lambda}(KVA^*VK)$$

$$KV(BA)^*VK = (KVA^*VK)(KVB^*VK)$$

Now if  $BA(KVA^*VK) = CA(KVA^*VK)$  then by (1)

$$BA(KVA^*VK) - CA(KVA^*VK) = 0$$

$$\Rightarrow (BA(KVA^*VK) - CA(KVA^*VK))(KV(B-C)^*VK) = 0$$

$$\Rightarrow (BA - CA)(KV(BA - CA)^*VK) = 0$$

$$\Rightarrow (BA - CA) = 0$$

$$\Rightarrow BA = CA$$

Therefore  $BA(KVA^*VK) = CA(KVA^*VK) \Rightarrow BA = CA \rightarrow (2)$

Similarly,  $B(KVA^*VK)A = C(KVA^*VK)A$

$$\Rightarrow B(KVA^*VK) = C(KVA^*VK) \rightarrow (3)$$

**Definition 1.1:** [3]

A Matrix  $A \in C_{n \times n}$  is said to be secondary k-normal (s-k normal) if

$$A(KVA^*VK) = (KVA^*VK)A$$

**Example 1.2:**

$$A = \begin{pmatrix} i & 2 & 3 & 4 \\ 4 & -i & 2 & 3 \\ 3 & 4 & i & 2 \\ 2 & 3 & 4 & -i \end{pmatrix} \text{ is a s-k normal matrix for } k=(1,3),(2,4) \text{ the permutation matrix be}$$

$$K = \begin{pmatrix} 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \end{pmatrix} \text{ and } V = \begin{pmatrix} 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \\ 1 & 0 & 0 & 0 \end{pmatrix}$$

**Definition 1.3:**

A matrix  $A \in C_{n \times n}$  is said to be secondary k-unitary (s-k unitary) if

$$A(KVA^*VK) = (KVA^*VK)A = I$$

**Example 1.4:**

$$A = \begin{pmatrix} i & -1 & 1 & 0 \\ -1 & -i & 0 & 1 \\ 1 & 0 & i & -1 \\ 0 & 1 & -1 & -i \end{pmatrix} \text{ is a s-k unitary matrix}$$

**Section 2: Secondary k - Generalized inverses of a matrix****Theorem 2.1:**

For any  $A \in C_{n \times n}$ , the four equations

$$AXA = A \rightarrow (4)$$

$$XAX = X \rightarrow (5)$$

$$KV(AK)^*VK = AK \rightarrow (6)$$

$$KV(XA)^*VK = XA \quad \rightarrow (7)$$

have a unique solution for any  $A \in C_{n \times n}$ .

**Proof:** First, we shall show that equations (5) & (6) are equivalent to the single equation

$$XKV(AX)^*VK = X \quad \rightarrow (8)$$

From equations (5) and (6), (7) follows, since it is merely (6) substituted in (5). Conversely, equation

(8) implies 
$$AXKV(AX)^*VK = AX$$

Since the left hand side is s-k hermitian, (6) follows. By substituting (6) in (8), we get  $XAX = X$  which is actually (5). Therefore (5) and (7) are equivalent to (8). Similarly, (4) & (7) are equivalent to the equation

$$XA(KVA^*VK) = KVA^*VK \rightarrow (9)$$

Thus to find a solution for the given set of equations, it is enough to find an X satisfying (8) & (9). Now the expressions  $((KVA^*VK)A), ((KVA^*VK)A)^2, ((KVA^*VK)A)^3 \dots$  cannot all be linearly independent (i.e) there exists a relation

$$\lambda_1((KVA^*VK)A) + \lambda_2((KVA^*VK)A)^2 + \dots + \lambda_k((KVA^*VK)A)^k = 0 \rightarrow (10)$$

Where  $\lambda_1, \lambda_2, \dots, \lambda_k$  are not all zero. Let  $\lambda_r$  be the first non zero  $\lambda$ . (i.e)  $\lambda_1 = \lambda_2 = \dots = \lambda_{r-1} = 0$ .

Therefore (10) implies that

$$\lambda_r((KVA^*VK)A)^r = -\left\{ \lambda_{r+1}((KVA^*VK)A)^{r+1} + \dots + \lambda_m((KVA^*VK)A)^m \right\}$$

$$\text{If we take } B = -\lambda_r^{-1} \left\{ \lambda_{r+1}I + \lambda_{r+2}((KVA^*VK)A) + \dots + \lambda_m((KVA^*VK)A)^{m-r-1} \right\}$$

Then

$$B((KVA^*VK)A)^{r+1} = -\lambda_r^{-1} \left\{ \lambda_{r+1}((KVA^*VK)A)^{r+1} + \dots + \lambda_m((KVA^*VK)A)^m \right\}$$

$B((KVA^*VK)A)^{r+1} = ((KVA^*VK)A)^r$ . By using (2) & (3) repeatedly, we get

$$B(KVA^*VK)A(KVA^*VK) = KVA^*VK \quad \rightarrow (11)$$

Now if we take  $X = B(KVA^*VK)$  then (11) implies that this X satisfies (9)

implies (7), we have  $(KV(XA)^*VK)(KVA^*VK) = KVA^*VK$

$$\Rightarrow B(KV(XA)^*VK)(KVA^*VK) = B(KVA^*VK)$$

Therefore  $X = B(KVA^*VK)$  satisfies (8). Thus  $X = B(KVA^*VK)$  is a solution for the given set of equations.

Now let us prove that this X is unique. Suppose that X and Y satisfy (8) and (9). Then by substituting (7) in (5) and (6) in (4), we obtain

$$(KV(XA)^*VK)X = X \text{ and } (KV(AX)^*VK)A = A$$

$$\text{Also, } Y = (KV(YA)^*VK)Y \text{ and } KVA^*VK = (KVA^*VK)AY$$

$$\text{Now } X = X(KVX^*VK)(KVA^*VK)$$

$$= X(KVX^*VK)(KVA^*VK)AY$$

$$= X(KV(AX)^*VK)AY$$

$$= XAY$$

$$= XA(KV(YA)^*VK)Y$$

$$= XA(KVA^*VK)(KVY^*VK)Y$$

$$= (KVA^*VK)(KVY^*VK)Y$$

$$= (KV(YA)^*VK)Y$$

$$X = Y$$

Therefore X is unique.

**Definition 2.2:** Let  $A \in C_{n \times n}$ . The unique solution of (4), (5), (6) and (7) is called secondary k-

generalized inverse of A and is written as  $A^{\dagger sk}$ .

**Example 2.3:**

$$\text{If } A = \begin{pmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{pmatrix} \text{ then } A^{\dagger sk} = \begin{pmatrix} 1/9 & 1/9 & 1/9 \\ 1/9 & 1/9 & 1/9 \\ 1/9 & 1/9 & 1/9 \end{pmatrix}$$

**Note 2.4:** By using (7) in (5), (6) in (4) and from (8) and (9) we obtain

$$\left. \begin{aligned} A^{\dagger sk} (KV(A^{\dagger sk})^* VK)(KVA^* VK) &= A^{\dagger sk} = (KVA^* VK)(KV(A^{\dagger sk})^* VK)A^{\dagger sk} \\ A^{\dagger sk} A(KVA^* VK) &= (KVA^* VK) = (KVA^* VK)AA^{\dagger sk} \end{aligned} \right\} \rightarrow (12)$$

If  $\lambda$  is a scalar, then  $\lambda^{\dagger sk}$  means  $\lambda^{-1}$  when  $\lambda \neq 0$  and zero when  $\lambda = 0$ .

### Section3: Secondary- k-generalized inverse of s-k normal matrices.

In this paper, characterizations of secondary k-generalized inverse (s-k-g) inverse of a matrix are obtained s-k-g inverse of s-k-normal matrices are discussed .s-k hermitian matrices are defined and the condition for s-k normal matrices to be diagonal is investigated.

**Theorem 3.1:** For  $A \in C_{n \times n}$ .

$$(i) \left( A^{\dagger sk} \right)^{\dagger sk} = A \quad (ii) \left( KV(A^*)^{\dagger sk} VK \right) = \left( KV \left( A^{\dagger sk} \right)^* VK \right)$$

$$(iii) \text{ If } A \text{ is non singular, then } A^{\dagger sk} = A^{-1} \quad (iv) (\lambda A)^{\dagger sk} = \lambda^{\dagger sk} A^{\dagger sk}$$

$$(v) ((KVA^* VK)A)^{\dagger sk} = A^{\dagger sk} (KVA^{\dagger sk} VK)^*$$

**Proof:** Let  $A \in C_{n \times n}$ .

(i) By the definition of s-k-g inverse, we have

$$A \dagger_{sk} A A \dagger_{sk} = A \dagger_{sk} \quad \text{and} \quad A \dagger_{sk} (A \dagger_{sk}) \dagger_{sk} A \dagger_{sk} = A \dagger_{sk}$$

These two equations imply that

$$A \dagger_{sk} \dagger_{sk} = A$$

(ii) From the definition of  $A \dagger_{sk}$ , we have  $AA \dagger_{sk} A = A$

$$\Rightarrow (KVA^*VK)(KV(A \dagger_{sk})^*VK)(KVA^*VK) = KVA^*VK$$

$$\text{Also } (KVA^*VK)(KV(A^*) \dagger_{sk} VK)(KVA^*VK) = KVA^*VK$$

From these two equations, we have

$$(KV(A \dagger_{sk})^*VK) = KV(A^*) \dagger_{sk} VK$$

(iii) Since A is non singular,  $A^{-1}$  exists

$$\text{Now } AA \dagger_{sk} A = A \quad (\text{By definition of } \dagger_{sk})$$

Pre multiplying & post multiplying by  $A^{-1}$  we have

$$A \dagger_{sk} = A^{-1}$$

(iv) The equations,  $AA \dagger_{sk} A = A$  and

$$(\lambda A)(\lambda A) \dagger_{sk} (\lambda A) = (\lambda A) \quad \text{imply that}$$

$$\lambda(\lambda A) \dagger_{sk} = A \dagger_{sk}$$

$$\Rightarrow (\lambda A)^{\dagger_{sk}} = \lambda^{\dagger_{sk}} A^{\dagger_{sk}} \text{ where } \lambda^{\dagger_{sk}} = \lambda^{-1}$$

(v) from (12) we have,

$$A^{\dagger_{sk}} (KV(A^{\dagger_{sk}})^* VK)(KVA^* VK) = A^{\dagger_{sk}}$$

$$\text{Also } AA^{\dagger_{sk}}A = A.$$

$$\text{Therefore } AA^{\dagger_{sk}}(KV(A^{\dagger_{sk}})^* VK)(KVA^* VK)A = A$$

Substitute this in the right hand side of the defining relation, we get

$$((KVA^* VK)A)^{\dagger_{sk}} = A^{\dagger_{sk}}(KV(A^{\dagger_{sk}})^* VK)$$

**Theorem 3.2:** A necessary and sufficient condition for the equation  $AXB = D$  to have a solution is

$$AA^{\dagger_{sk}}DB^{\dagger_{sk}}B = D, \text{ in which case the general solution is}$$

$$X = A^{\dagger_{sk}}DB^{\dagger_{sk}} + Y - A^{\dagger_{sk}}AYBB^{\dagger_{sk}}, \text{ where } Y \text{ is arbitrary.}$$

**Proof:** Let us assume that  $X$  satisfies the equation  $AXB = D$ , then  $D = AXB$

$$= AA^{\dagger_{sk}}AXB^{\dagger_{sk}}B = AA^{\dagger_{sk}}DB^{\dagger_{sk}}B \text{ (By the definition of } \dagger_{sk} \text{)}$$

Conversely if  $D = AA^{\dagger_{sk}}DB^{\dagger_{sk}}B$ , then  $X = A^{\dagger_{sk}}DB^{\dagger_{sk}}$ , then it is a particular solution of  $AXB = D$ . since  $AXB = AA^{\dagger_{sk}}DB^{\dagger_{sk}}B = D$ .

If  $Y \in C_{n \times n}$ , then any expression of the form  $X = A^{\dagger_{sk}}DB^{\dagger_{sk}} + Y - A^{\dagger_{sk}}AYBB^{\dagger_{sk}}$  is a solution of  $AXB = D$ . and conversely, if  $X$  is a solution  $AXB = D$ , then  $X = A^{\dagger_{sk}}DB^{\dagger_{sk}} + X - A^{\dagger_{sk}}AXB^{\dagger_{sk}}B$  satisfies  $AXB = D$ . Hence the theorem.

**Theorem 3.3:** The matrix equations  $AX = B$  and  $XD = E$  have a common solution if and only if each equation has a solution and  $AE = BD$ .



**Proof:** It is easy to see that the conditions is necessary, conversely  $A \dagger_{sk} B$  and  $ED \dagger_{sk}$  are solutions of  $AX=B$  and  $XD=E$  and hence  $AA \dagger_{sk} B=B$  and  $ED \dagger_{sk} D=E$ . Also  $AE=BD$ . By using these facts it can be prove that  $X = A \dagger_{sk} B + ED \dagger_{sk} - A \dagger_{sk} AED \dagger_{sk}$  is a common solution of the given equations.

**Definition 3.4:** A matrix  $E \in C_{n \times n}$  is said to be secondary-k hermitian idempotent matrix (s-k. h.i) if  $E(KVE^*VK) = E$  (i.e)  $E = KVE^*VK$  and  $E^2 = E$ .

**Theorem 3.5:** (i)  $A \dagger_{sk} A$ ,  $AA \dagger_{sk}$ ,  $1 - A \dagger_{sk} A$ ,  $1 - AA \dagger_{sk}$  are all the s-k hermitian idempotent.

(ii)  $J$  is idempotent  $\Leftrightarrow$  there exist s-k hermitian idempotent's  $E$  and  $F$  such that  $J = (FE) \dagger_{sk}$  in which case  $J = EJF$ .

**Proof:** Proof of (i) is obvious. If  $J$  is idempotent then  $J^2=J$ . By (i) of theorem (3.1),

$J = \left\{ (J \dagger_{sk} J)(JJ \dagger_{sk}) \right\} \dagger_{sk}$ . Now if we take  $E = JJ \dagger_{sk}$  and  $F = J \dagger_{sk} J$  they will satisfy our

requirements conversely if  $J = (FE) \dagger_{sk}$  then  $J=EFPEF$  where

$P = (KV((FE) \dagger_{sk})^*VK)(FE) \dagger_{sk} (KV((FE) \dagger_{sk})^*VK)$ . Therefore  $J=EJF$  and hence

$J^2 = E(FE) \dagger_{sk} FE(FE) \dagger_{sk} F = E(FE) \dagger_{sk} F = J$ . Hence  $J$  is idempotent.

**Note 3.6:**(i) s-k hermitian idempotent matrices are s-k normal matrices.

(ii) The s-k-g inverse of an s-k hermitian idempotent matrix is also s-k hermitian idempotent matrix.

**Definition 3.7:** For any square matrix  $A$  there exists a unique set of matrices  $J_\lambda$  defined for each complex number  $\lambda$  such that

$$J_\lambda J_\mu = \delta_{\lambda\mu} J_\lambda \quad \rightarrow (13)$$

$$\sum J_\lambda = 1 \quad \rightarrow (14)$$

$$AJ_{\lambda} = J_{\lambda}A \quad \rightarrow (15)$$

$$(A - \lambda I)J_{\lambda} \text{ is nilpotent } \rightarrow (16)$$

Then the non zero  $J_{\lambda}$ 's are called the principal idempotent elements of A.

**Theorem 3.8:** If  $E_{\lambda} = 1 - \{(A - \lambda I)^n\}^{\dagger sk} (A - \lambda I)^n$  and  $F_{\lambda} = 1 - (A - \lambda I)^n \{(A - \lambda I)^n\}^{\dagger sk}$ ,

where n is sufficiently large, then the principal idempotent element of A are given by

$$J_{\lambda} = \{F_{\lambda} E_{\lambda}\}^{\dagger sk} \text{ and n can be taken as unity iff A is diagonal.}$$

**Proof:** Assume that A is diagonal.

$$\text{Let } E_{\lambda} = 1 - \{(A - \lambda I)\}^{\dagger sk} (A - \lambda I), \text{ and } F_{\lambda} = 1 - (A - \lambda I)\{(A - \lambda I)\}^{\dagger sk}$$

Then by 3.5(i)  $E_{\lambda}$  and  $F_{\lambda}$  are s-k hermitian idempotent matrices. If  $\lambda$  is not an eigen value of A, then  $|A - \lambda I| \neq 0$  and hence  $F_{\lambda}$  and  $E_{\lambda}$  are zero by (iii) of theorem(3.1). Clearly,  $(A - \mu I)E_{\mu} = 0$  and  $F_{\lambda}(A - \lambda I) = 0 \rightarrow (17)$

$$\text{Therefore } \mu F_{\lambda} E_{\mu} = F_{\lambda} A E_{\mu} = \lambda F_{\lambda} E_{\mu}$$

$$\text{Hence } F_{\lambda} E_{\mu} = 0 \quad \text{if } \lambda \neq \mu \rightarrow (18)$$

Now if we take  $J_{\lambda} = \{F_{\lambda} E_{\lambda}\}^{\dagger sk}$  then by (ii) of theorem (3.5),

$$J_{\lambda} = E_{\lambda} \{F_{\lambda} E_{\lambda}\}^{\dagger sk} F_{\lambda} \quad \rightarrow (19),$$

Now(18) implies  $J_{\lambda} J_{\mu} = \delta_{\lambda\mu} J_{\lambda}$ . Also by (18),  $F_{\lambda} J_{\mu} E_{\gamma} = \delta_{\lambda\mu} \delta_{\mu\gamma} F_{\lambda} E_{\lambda} \rightarrow (20)$  If  $Z_{\alpha}$

is an eigen vector of A corresponding to the eigen value  $\alpha$  then  $E_{\alpha} Z_{\alpha} = Z_{\alpha}$ . As A is diagonal, any column vector X conformable with A is expressible as a sum of eigen vectors (i.e) it is expressible in the form  $X = \sum E_{\lambda} X_{\lambda}$ . This is a finite sum over all complex  $\lambda$ .

Similarly, if  $Y^*$  is conformable with A, it is expressible as  $Y^* = \sum Y_\lambda^* F_\lambda$  Now by

$$\text{equations (18) and (20)} \quad Y^* (\sum J_\mu) X = (\sum Y_\lambda^* F_\lambda) (\sum J_\mu) (\sum E_\gamma X_\lambda)$$

$$= \sum Y_\lambda^* F_\lambda E_\lambda X_\lambda$$

$$= Y^* (\sum J_\mu) X = Y^* X$$

$$\Rightarrow Y^* (\sum J_\mu - I) X = 0$$

$$\Rightarrow \sum J_\mu = I$$

$$\text{Also(17) and (19) lead to } AJ_\lambda = \lambda J_\lambda = J_\lambda A \rightarrow (21)$$

This implies  $(A - \lambda I)J_\lambda$  is nilpotent and (15) and (16) are satisfied.

$$\text{Moreover } A = \sum \lambda J_\lambda \rightarrow (22)$$

Conversely if  $\sum J_\lambda = I$  and A is not diagonalizable (n=1) then  $X = \sum J_\lambda X$  gives X as a sum of eigen vectors of A, since (21) was derived without assuming the diagonality of A. If A is not diagonalizable. It seems more convenient simply to prove that for any set of  $J_\lambda$ 's satisfying (19), (20), (21) & (22) each  $J_\lambda = (F_\lambda E_\lambda)^\dagger_{sk}$  where  $F_\lambda$  and  $E_\lambda$  are defined as in the theorem.

If the  $J_\lambda$ 's satisfy (13), (14), (15) & (16)  $\sum J_\lambda = I$  and

$$(A - \lambda I)^n J_\lambda = 0 = J_\lambda (A - \lambda I)^n \rightarrow (23)$$

Which comes by using the fact that  $(A - \lambda I)J_\lambda$  is nilpotent, where n is sufficiently large.

From (23) and the definition of  $E_\lambda$  and  $F_\lambda$ , we have

$$E_\lambda F_\lambda = J_\lambda = J_\lambda F_\lambda \rightarrow (24)$$

By using Euclid's algorithm, there exist P and Q which are polynomials in A such that

$$I = (A - \lambda I)^n P + Q(A - \mu I)^n \quad \text{if } \lambda \neq \mu.$$

$$\text{Now } F_\lambda (A - \lambda I)^n = 0 = (A - \mu I)^n E_\mu. \quad \text{Hence } F_\lambda E_\mu = 0 \text{ if } \lambda \neq \mu$$

$$\text{From (24) } F_\lambda J_\mu = 0 = J_\lambda E_\mu \text{ if } \lambda \neq \mu. \quad \text{Since } \sum J_\mu = I, \text{ we get}$$

$$F_\lambda J_\lambda = F_\lambda \text{ and } J_\lambda E_\lambda = E_\lambda \rightarrow (25) \text{ using (24) and (25) it is easy to see that}$$

$$\{F_\lambda E_\lambda\}^{\dagger s-k} = J_\lambda. \quad \text{Hence the theorem}$$

**Theorem 3.9:** If A is s-k normal, it is diagonalizable and its principal idempotent elements are s-k hermitian.

**Proof:** If A is s-k normal then  $(A - \lambda I)$  is s-k normal. By using (viii) of the theorem (3.1) in the

definition of  $E_\lambda$  and  $F_\lambda$  of theorem (3.8) we obtain  $E_\lambda = 1 - \{(A - \lambda I)\}^{\dagger sk} (A - \lambda I)$  and

$$F_\lambda = 1 - (A - \lambda I)\{(A - \lambda I)\}^{\dagger sk}$$

Hence A is diagonalizable. since  $E_\lambda = F_\lambda, J_\lambda = E_\lambda$  is s-k hermitian.

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