

## k - Symmetric Doubly Stochastic, s - Symmetric Doubly Stochastic and s – k - Symmetric Doubly Stochastic Matrices

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**ABSTRACT:** The basic concepts and theorems of k-symmetric doubly stochastic s-symmetric doubly stochastic and s-k-symmetric doubly stochastic matrices are introduced with examples.

**KEY WORDS:** k-symmetric doubly stochastic matrix, s-symmetric doubly stochastic matrix and s-k-symmetric doubly stochastic matrix.

**AMS CLASSIFICATIONS:** 15A51, 15B99

### I. INTRODUCTION

We have already seen the concept of symmetric doubly stochastic matrices. In this paper the symmetric doubly stochastic matrix is developed in real matrices. Recently Hill and Waters[2] have developed a theory of k-real matrices as a generalization of s-real matrices. Ann Lee[1] has initiated the study of secondary symmetric matrices, that is matrices whose entries are symmetric about the secondary diagonal. Ann Lee[1] has shown that the matrix A, the usual transpose  $A^T$  and secondary transpose  $A^S$  are related as  $A^S = VA^T V$  and  $A^T = VA^S V$  where V is a permutation matrix with units in the secondary diagonal.

### II. PRELIMINARIES AND NOTATIONS

$A^T$  - Transpose of A

Let k be a fixed product of disjoint transpositions in  $S_n$  and 'K' be the permutation matrix associated with k. Clearly K satisfies the following properties.  $K^2 = I, K^T = K$ .

### III. DEFINITIONS AND THEOREMS

**DEFINITION: 1**

A matrix  $A \in R^{n \times n}$  is said to be symmetric doubly stochastic matrix if  $A = A^T$  and  $\sum_{i=1}^n a_{ij} = 1, j = 1, 2, \dots, n$  and  $\sum_{j=1}^n a_{ij} = 1, i = 1, 2, \dots, n$  and all  $a_{ij} \geq 0$ .

If A is doubly stochastic and also symmetric then it is called a symmetric doubly stochastic matrix.

**DEFINITION: 2**

A matrix  $A \in R^{n \times n}$  is said to be k-symmetric doubly stochastic matrix if  $A = K A^T K$

**THEOREM: 1**

Let  $A \in R^{n \times n}$  is k-symmetric doubly stochastic matrix then  $A = K A^T K$ .

**Proof:**

$$\begin{aligned} K A^T K &= K A K \text{ where } A^T = A \\ &= A K K \text{ where } A K = K A \\ &= A K^2 = A \text{ where } K^2 = I \end{aligned}$$

**THEOREM: 2**

Let  $A^T \in R^{n \times n}$  is k-symmetric doubly stochastic matrix then  $A^T = K A K$ .

**Proof:**

$$\begin{aligned} K A K &= K A^T K \text{ where } A = A^T \\ &= A^T K K \text{ where } K A^T = A^T K \\ &= A^T K^2 = A^T \text{ where } K^2 = I \end{aligned}$$

**THEOREM: 3**

Let  $A, B \in \mathbb{R}^{n \times n}$  is  $k$ -symmetric doubly stochastic matrix then  $\frac{1}{2}(A + B)$  is  $k$ -symmetric doubly stochastic matrix.

**Proof:**

Let  $A$  and  $B$  are  $k$ -symmetric doubly stochastic matrix if  $A = K A^T K$  and  $B = K B^T K$ .

To prove  $\frac{1}{2}(A + B)$  is  $k$ -symmetric doubly stochastic matrix we will show that

$$\frac{1}{2}(A + B) = K \frac{1}{2}(A + B)^T K$$

$$\begin{aligned} \text{Now } K \frac{1}{2}(A + B)^T K &= K \frac{1}{2}(A^T + B^T) K = \frac{1}{2} K (A^T + B^T) K = \frac{1}{2} (K A^T + K B^T) K = \frac{1}{2} (K A^T K + K B^T K) \\ &= \frac{1}{2}(A + B) \text{ where } K A^T K = A \text{ and } K B^T K = B \end{aligned}$$

**THEOREM: 4**

Any  $k$ -symmetric doubly stochastic matrix can be represent as sum of  $k$ -symmetric doubly stochastic matrix and skew  $k$ -symmetric doubly stochastic matrix.

**Proof:**

To prove that  $\frac{1}{2}(A + K A^T K)$  and  $\frac{1}{2}(A - K A^T K)$  are  $k$ -symmetric doubly stochastic matrices the we will show that  $\frac{1}{2}(A + K A^T K) = K \frac{1}{2}(A + K A^T K)^T K$  and  $\frac{1}{2}(A - K A^T K) = K \frac{1}{2}(A - K A^T K)^T K$ .

$$K \frac{1}{2}(A + K A^T K)^T K = \frac{1}{2}(A + K A^T K) \text{ using theorem 3 and } K \frac{1}{2}(A - K A^T K)^T K = \frac{1}{2}(A - K A^T K).$$

Then  $\frac{1}{2}(A + K A^T K) + \frac{1}{2}(A - K A^T K) = 2A/2 = A$ . Hence the theorem is proved.

**THEOREM: 5**

If  $A$  and  $B$  are  $k$ -symmetric doubly stochastic matrices then  $AB$  is also  $k$ -symmetric doubly stochastic matrix.

**Proof:**

Let  $A$  and  $B$  are  $k$ -symmetric doubly stochastic matrix if  $A = K A^T K$  and  $B = K B^T K$ .

Since  $A^T$  and  $B^T$  are also  $k$ -symmetric doubly stochastic matrices then  $A^T = K A K$  and  $B^T = K B K$ .

To prove  $AB$  is  $k$ -symmetric doubly stochastic matrix we will show that

$$AB = K (A B)^T K$$

$$\begin{aligned} \text{Now } K (A B)^T K &= K B^T A^T K = K (K B K) (K A K) K \text{ where } A^T = K A K \text{ and } B^T = K B K. \\ &= K^2 B K^2 A K^2 = B A \text{ where } K^2 = I \\ &= AB \text{ where } BA = AB \end{aligned}$$

**THEOREM: 6**

If  $A$  and  $B$  are  $k$ -symmetric doubly stochastic matrices and  $K$  is the permutation matrix,  $k = \{(1), (2\ 3)\}$  then  $KA$  is also  $k$ -symmetric doubly stochastic matrix.

**Proof:**

Let  $A$  and  $B$  are  $k$ -symmetric doubly stochastic matrix if  $A = K A^T K$  and  $B = K B^T K$ .

Since  $A^T$  and  $B^T$  are also  $k$ -symmetric doubly stochastic matrices then  $A^T = K A K$  and  $B^T = K B K$ .

To prove  $KA$  is  $k$ -symmetric doubly stochastic matrix we will show that  $KA = K (KA)^T K$

$$\begin{aligned} \text{Now } K (KA)^T K &= K (A^T K^T) K = K A^T K^T K = K A^T \text{ where } K^T K = I. \\ &= KA \text{ where } KA^T = KA \end{aligned}$$

**RESULT:**

For  $A \in \mathbb{R}^{n \times n}$  is symmetric doubly stochastic matrices for the following are holds.

- [1]  $A = K A^T K$
- [2]  $KA$  is symmetric doubly stochastic matrix.
- [3]  $AK$  is symmetric doubly stochastic matrix.

**Example:**

$$A = \begin{pmatrix} 0 & 1/2 & 1/2 \\ 1/2 & 0 & 1/2 \\ 1/2 & 1/2 & 0 \end{pmatrix} \quad A^T = \begin{pmatrix} 0 & 1/2 & 1/2 \\ 1/2 & 0 & 1/2 \\ 1/2 & 1/2 & 0 \end{pmatrix} \quad \text{and } k = (1) (2\ 3) = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{pmatrix}$$

$$(i) \quad K A^T K = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{pmatrix} \begin{pmatrix} 0 & 1/2 & 1/2 \\ 1/2 & 0 & 1/2 \\ 1/2 & 1/2 & 0 \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{pmatrix} = A$$

Similarly  $KAK = A^T$

$$(ii) \quad KA = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{pmatrix} \begin{pmatrix} 0 & 1/2 & 1/2 \\ 1/2 & 0 & 1/2 \\ 1/2 & 1/2 & 0 \end{pmatrix} = \begin{pmatrix} 0 & 1/2 & 1/2 \\ 1/2 & 1/2 & 0 \\ 1/2 & 0 & 1/2 \end{pmatrix} = (KA)^T$$

$\Rightarrow KA$  is symmetric doubly stochastic matrix.

Similarly  $KA^T$  is also symmetric doubly stochastic matrix.

$$(iii) \quad AK = \begin{pmatrix} 0 & 1/2 & 1/2 \\ 1/2 & 0 & 1/2 \\ 1/2 & 1/2 & 0 \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{pmatrix} = \begin{pmatrix} 0 & 1/2 & 1/2 \\ 1/2 & 1/2 & 0 \\ 1/2 & 0 & 1/2 \end{pmatrix} = (AK)^T$$

$\Rightarrow AK$  is symmetric doubly stochastic matrix.

Similarly  $A^T K$  is also symmetric doubly stochastic matrix.

**DEFINITION: 3**

A matrix  $A \in R^{n \times n}$  is said to be s-symmetric doubly stochastic matrix if  $A^S = V A^T V$  where  $V$  is a permutation matrix with units in the secondary diagonal.

**THEOREM: 7**

Let  $A \in R^{n \times n}$  is s-symmetric doubly stochastic matrix then  $A^S = V A^T V$ .

**Proof:**

$$V A^T V = V (V A^S V) V = V^2 A^S V^2 = A^S \text{ where } V^2 = I$$

**THEOREM: 8**

Let  $A^T \in R^{n \times n}$  is s-symmetric doubly stochastic matrix then  $A^T = V A^S V$ .

**Proof:**

$$V A^S V = V (V A^T V) V = V^2 A^T V^2 = A^T \text{ where } V^2 = I$$

**THEOREM: 9**

Let  $A, B \in R^{n \times n}$  is s-symmetric doubly stochastic matrix then  $\frac{1}{2}(A + B)$  is s-symmetric doubly stochastic matrix.

**Proof:**

Let  $A$  and  $B$  are s-symmetric doubly stochastic matrices if  $A^S = V A^T V$  and  $B^S = V B^T V$ .

To prove  $\frac{1}{2}(A + B)$  is s-symmetric doubly stochastic matrix we will show that

$$\frac{1}{2}(A + B)^S = V \frac{1}{2}(A + B)^T V$$

$$\begin{aligned} \text{Now } V \frac{1}{2}(A + B)^T V &= V \frac{1}{2}(A^T + B^T) V = \frac{1}{2} V(A^T + B^T) V = \frac{1}{2} (V A^T + V B^T) V = \frac{1}{2} (V A^T V + V B^T V) \\ &= \frac{1}{2}(A^S + B^S) \text{ where } V A^T V = A^S \text{ and } V B^T V = B^S \\ &= \frac{1}{2}(A + B)^S \end{aligned}$$

**THEOREM: 10**

If  $A$  and  $B$  are s-symmetric doubly stochastic matrices then  $AB$  is also s-symmetric doubly stochastic matrix.

**Proof:**

Let  $A$  and  $B$  are s-symmetric doubly stochastic matrices if  $A^S = V A^T V$  and  $B^S = V B^T V$ .

Since  $A^T$  and  $B^T$  are also s-symmetric doubly stochastic matrices then  $A^T = V A^S V$  and  $B^T = V B^S V$ .

To prove  $AB$  is s-symmetric doubly stochastic matrix we will show that

$$(AB)^S = V(AB)^T V$$

$$\begin{aligned} \text{Now } V(AB)^T V &= V B^T A^T V = V(V B^S V)(V A^S V) V \quad \text{where } A^T = V A^S V \text{ and } B^T = V B^S V \\ &= V^2 B^S V^2 A^S V^2 = B^S A^S \text{ where } V^2 = I \\ &= (AB)^S \end{aligned}$$

**THEOREM: 11**

If A is s-symmetric doubly stochastic matrix and V is a permutation matrix with units in the secondary diagonal then VA is also s-symmetric doubly stochastic matrix.

**Proof:**

Let A is s-symmetric doubly stochastic matrices if  $A^S = V A^T V$ . Since  $A^T$  is s-symmetric doubly stochastic matrices then  $A^T = V A^S V$ . To prove VA is s-symmetric doubly stochastic matrix we will show that

$$\begin{aligned} \text{Now } V(VA)^T V &= V(A^T V^T) V = V(VA^S V) V^2 \\ &= A^S V^S \text{ where } V^2 = I \\ &= (VA)^S \end{aligned}$$

**RESULT:**

For  $A \in R^{n \times n}$  is s-symmetric doubly stochastic matrix for the following are holds.

- [1]  $A^S = V A^T V$
- [2] VA is symmetric doubly stochastic matrix.
- [3] AV is symmetric doubly stochastic matrix.

**Example:**

$$A = \begin{pmatrix} 0 & 1/2 & 1/2 \\ 1/2 & 0 & 1/2 \\ 1/2 & 1/2 & 0 \end{pmatrix} \quad A^T = A^S = \begin{pmatrix} 0 & 1/2 & 1/2 \\ 1/2 & 0 & 1/2 \\ 1/2 & 1/2 & 0 \end{pmatrix} \quad V = \begin{pmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{pmatrix}$$

$$(i) \quad V A^T V = \begin{pmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{pmatrix} \begin{pmatrix} 0 & 1/2 & 1/2 \\ 1/2 & 0 & 1/2 \\ 1/2 & 1/2 & 0 \end{pmatrix} \begin{pmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{pmatrix} = A^S$$

Similarly  $VA^{SV} = A^T$

$$(ii) \quad VA = \begin{pmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{pmatrix} \begin{pmatrix} 0 & 1/2 & 1/2 \\ 1/2 & 0 & 1/2 \\ 1/2 & 1/2 & 0 \end{pmatrix} = \begin{pmatrix} 1/2 & 1/2 & 0 \\ 1/2 & 0 & 1/2 \\ 0 & 1/2 & 1/2 \end{pmatrix} = (VA)^T$$

$\Rightarrow$  VA is symmetric doubly stochastic matrix.

Similarly  $VA^T$  is also symmetric doubly stochastic matrix.

$$(iii) \quad AV = \begin{pmatrix} 0 & 1/2 & 1/2 \\ 1/2 & 0 & 1/2 \\ 1/2 & 1/2 & 0 \end{pmatrix} \begin{pmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{pmatrix} = \begin{pmatrix} 1/2 & 1/2 & 0 \\ 1/2 & 0 & 1/2 \\ 0 & 1/2 & 1/2 \end{pmatrix} = (AV)^T$$

$\Rightarrow$  AV is symmetric doubly stochastic matrix.

Similarly  $A^T V$  is also symmetric doubly stochastic matrix.

**DEFINITION: 4**

A matrix  $A \in R^{n \times n}$  is said to be s-k-symmetric doubly stochastic matrix if

- [1]  $A = KVA^T VK$
- [2]  $A^T = KVAVK$
- [3]  $A = VKA^T KV$
- [4]  $A^T = VKAKV$

Where V is a permutation matrix with units in the secondary diagonal and K is a permutation matrix and  $k = \{(1) (2 \ 3)\}$ .

**THEOREM: 12**

Let  $A \in R^{n \times n}$  is s-k-symmetric doubly stochastic matrix then

- [1]  $A^S = KVA^T VK$
- [2]  $A^T = KVA^S VK$
- [3]  $A^S = VKA^T KV$
- [4]  $A^T = VKA^S KV$

**Proof:**

- (i)  $KVA^T VK = K(VA^T V)K$   
 $= KA^S K$  where  $VA^T V = A^S$   
 $= K(A^T)^T K = K A^T K$  where  $A^T = A$   
 $= A = A^S$  where  $KA^T K = A$  and  $A = A^S$
- (ii)  $KVA^S VK = K(VA^S V)K = K A^T K$  where  $VA^S V = A^T$   
 $= A$  where  $KA^T K = A$   
 $= A^T$  where  $A = A^T$
- (iii)  $VKA^T KV = V(KA^T K)V$   
 $= VAV$  where  $KA^T K = A$   
 $= VA^T V$  where  $A = A^T$   
 $= A^S$  where  $VA^T V = A^S$
- (iv)  $VKA^S KV = V(KA^S K)V = V(K(A^T)^T K)V$   
 $= VA^T V$  where  $K(A^T)^T K = A^T$   
 $= A^S$  where  $VA^T V = A^S$   
 $= (A^T)^T = A^T$  where  $A = A^T$

**THEOREM: 13**

Let  $A, B \in \mathbb{R}^{n \times n}$  is s-k-symmetric doubly stochastic matrix then  $\frac{1}{2}(A + B)$  is s-k-symmetric doubly stochastic matrix.

**Proof:**

Let A and B are s-k-symmetric doubly stochastic matrix if  $A = KV A^T VK$  and  $B = KVB^T VK$ .

To prove  $\frac{1}{2}(A + B)$  is s-k-symmetric doubly stochastic matrix we will show that

$$\frac{1}{2}(A + B) = KV \frac{1}{2}(A + B)^T VK$$

Now  $KV \frac{1}{2}(A + B)^T VK = K(V \frac{1}{2}(A + B)^T V)K = K \frac{1}{2}(A + B)^S K$  using theorem (9)  
 $= K \frac{1}{2}(A + B)^T K = \frac{1}{2}(A + B)$  using theorem (3)

**THEOREM: 14**

If A and B are s-k-symmetric doubly stochastic matrix then AB is also s-k-symmetric doubly stochastic matrix.

**Proof:**

Let A and B are s-k-symmetric doubly stochastic matrix if  $A = KV A^T VK$  and  $B = KVB^T VK$ .

Since  $A^T$  and  $B^T$  are also s-k-symmetric doubly stochastic matrices  $A^T = KVAVK$  and  $B^T = KVBVK$ .

To prove AB is s-k-symmetric doubly stochastic matrix we will show that

$$AB = KV(A B)^T VK$$

Now  $KV(A B)^T VK = K(V(A B)^T V)K = K(A B)^S K$  using theorem (10)  
 $= K(A B)^T K = AB$  using theorem (5)

**RESULT:**

For  $A \in \mathbb{R}^{n \times n}$  is s-k-symmetric doubly stochastic matrix the following are equivalent.

- (i)  $A = KVA^T VK$
- (ii)  $A^T = KVAVK$
- (iii)  $A = VKA^T KV$
- (iv)  $A^T = VKAKV$

**Example:**

$$A = \begin{pmatrix} 0 & 1/2 & 1/2 \\ 1/2 & 0 & 1/2 \\ 1/2 & 1/2 & 0 \end{pmatrix} \quad K = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{pmatrix} \quad V = \begin{pmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{pmatrix}$$

(i)  $KVA^T VK = \begin{pmatrix} 0 & 0 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix} \begin{pmatrix} 0 & 1/2 & 1/2 \\ 1/2 & 0 & 1/2 \\ 1/2 & 1/2 & 0 \end{pmatrix} \begin{pmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 0 & 0 \end{pmatrix} = A$

(ii)  $KVAVK = \begin{pmatrix} 0 & 0 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix} \begin{pmatrix} 0 & 1/2 & 1/2 \\ 1/2 & 0 & 1/2 \\ 1/2 & 1/2 & 0 \end{pmatrix} \begin{pmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 0 & 0 \end{pmatrix} = A^T$

$$(iii) \quad VKA^T KV = \begin{pmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 0 & 0 \end{pmatrix} \begin{pmatrix} 0 & 1/2 & 1/2 \\ 1/2 & 0 & 1/2 \\ 1/2 & 1/2 & 0 \end{pmatrix} \begin{pmatrix} 0 & 0 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix} = A$$

$$(iv) \quad VKAKV = \begin{pmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 0 & 0 \end{pmatrix} \begin{pmatrix} 0 & 1/2 & 1/2 \\ 1/2 & 0 & 1/2 \\ 1/2 & 1/2 & 0 \end{pmatrix} \begin{pmatrix} 0 & 0 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix} = A^T$$

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