



Some Inequality of Fuzzy Matrices

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Abstract: In this paper, the fuzzy matrix geometric mean is fuzzy concave. We complete this important fact with a reverse result. This follows from an interesting non-commutative extension of a classical reverse Cauchy Schwarz inequality in $M_n(F)$.

Keywords: Fuzzy matrix, determinant of a squarefuzzy matrix, , fuzzy norm, Cauchy Schwarz inequality

I. Introduction

The concept of fuzzy set was introduced by Zadeh [10] in 1965. R. Biswas [2] and A. M. K.H.kim and F.W.Roush [4] first tried to give a meaningful definition of Generalized fuzzy matrices and associated fuzzy norm function. Jian Miao Chen[3] introduced Fuzzy matrix partial ordering and generalized inverse. [1]Bertoliuzzaintrduced the properties ofdistributivity of t-norm and t-conorms.[5] Ragab.M. Z. and Emam E. G. introduced the concept of determinant and have the developed adjoint of a square fuzzy matrix. introduced On fuzzy 2-normed linear spaces. Meenakshi andCokilavany [6],[7,8] introduced the properties of fuzzy m-norm matrices and Binormed sequences in fuzzy matricesGani and Kalyani. Introduced the definition of positive Definite Matrices[2] R.Bhatia. Recently [9] ZHOU Min - na have introduced a definition of Characterizations of the Minus Ordering in Fuzzy Matrix Set^ In this paper, the concept of inequalityof fuzzy matrices on $M_n(F)$, the set of all fuzzy sets of $P^*(M_n(F))$, the standard Cauchy Schwarz inequality is proved. Moreover Cauchy Schwarz inequality of fuzzy matrices are established. Some equivalent conditions are also proved.

II. Preliminaries

We consider $F=[0,1]$ the fuzzy algebra with operation $[+,\cdot]$ and the standard order “ \leq ” where $a+b = \max\{a,b\}$, $a\cdot b = \min\{a,b\}$ for all a,b in F . F is a commutative semi-ring with additive and multiplicative identities 0 and 1 respectively. Let $M_{MN}(F)$ denote the set of all $m \times n$ fuzzy matrices over F . In short $M_n(F)$ is the set of all fuzzy matrices of order n . define ‘+’ and scalar multiplication in $M_n(F)$ as $A + B = [a_{ij} + b_{ij}]$, where $A = [a_{ij}]$ and $B = [b_{ij}]$ and $cA = [ca_{ij}]$, where c is in $[0,1]$, with these operations $M_n(F)$ forms a linear space.

Definition 2.1

An $m \times n$ matrix $A = [a_{ij}]$ whose components are in the unit interval $[0,1]$ is called a fuzzy matrix.

Definition 2.2

The determinant $|A|$ of an $n \times n$ fuzzy matrix A is defined as follows;

$$|A| = \sum_{\sigma \in S_n} a_{1\sigma(1)}a_{2\sigma(2)} \cdots a_{n\sigma(n)}$$

Where S_n denotes the symmetric group of all permutations of the indices $(1,2, \dots, n)$

Definition 2.3

A fuzzy matrix A is defined to be greater than B if $\|B\| \leq \|A\|$, A is strictly greater than B if $\|B\| < \|A\|$. We also say that B is smaller than A .

Definition 2.4

Define a mapping $d:M_n(F) \times M_n(F) \rightarrow [0,1]$ as $\|A + B\| = \det[A, B]$ for all A, B in $M_n(F)$.

Definition 2.5

A matrix A in $M_n(F)$ is called idempotentif $A^2 = A$

or $\|A^2\| = \det[A]$, where $A = [a_{ij}]$

Definition 2.6

Let A be in $M_n(F)$ and α be in $[0,1]$ such that $\|A\| = \alpha$ or $n(P_A) = \alpha$, then the pair (P_A, α) is called a fuzzy point in $M_n(F)$ and it is denoted by P_A^α . The dual fuzzy point P_A^α is the point with norm $(1-\alpha)$ denoted by $P^* = P^{1-\alpha}$.

Definition 2.7

The set of all fuzzy points in $M_n(F)$ is given by

$$P^*(M_n(F)) = \{P_A^\alpha \mid A \in M_n(F), \alpha \in [0,1]\}$$

In F we follow the usual \leq order relation correspondingly we define an order relation in $P^*(M_n(F))$

Definition 2.8

The Geometric mean \sqrt{ab} of two positive real numbers $a, b (0 \leq a, b \leq 1)$ is a convex operation. This paper equivalent to the Cauchy-Schwarz inequality. Let P_A, P_B, \dots, P_Z be $n \times n$ Fuzzy Matrices or operators on a n -dimensional space $P^*(M_n(F))$, for $P_A, P_B \in M_n(F)$ their geometric mean $P_A \# P_B$ is defined by two quite natural requirements.

1. $P_A P_B = P_B P_A$ implies $P_A \# P_B = \sqrt{P_A P_B}$
2. $(P_X^* P_A P_X) \# (P_X^* P_B P_X) = P_X^* (P_A \# P_B) P_X$

for any invertible P_X on $M_n(F)$.

$$\text{Then, } P_A \# P_B = P_A^{1/2} \left(I \# P_A^{-1/2} P_B P_A^{-1/2} \right) P_A^{1/2}$$

$$P_A \# P_B = P_A^{1/2} \left(P_A^{-1/2} P_B P_A^{-1/2} \right) P_A^{1/2} \quad \text{-----(1)}$$

So that $P_A \# P_B$ should be solution of $P_Z P_A^{-1} P_Z = P_B$, P_Z in $M_n(F)$ or equivalently to $P_Z P_B^{-1} P_Z = P_A$.

Hence, $P_A \# P_B$ can be defined (1) and $P_A \# P_B = P_B \# P_A$.

Since $P_A \# P_B$ is operator increasing. Since $f(t) = t^{1/2}$ is operator monotone.

Theorem 2.9

Let $P_A, P_B > 0$ in $M_n(F)$. Then $P_A \# P_B = \text{Max} \left\{ P_X \in [0,1] \mid \begin{pmatrix} P_A & P_X \\ P_X & P_B \end{pmatrix} \geq 0 \right\}$.

Proof:

$\begin{pmatrix} P_A & P_X \\ P_X & P_B \end{pmatrix} \geq 0$, if P_A, P_B, P_X in $M_n(F)$ where $P_X = P_A^{1/2} K P_B^{1/2}$ for some contraction

$K \in [0,1]$ and $n(K) \leq 1$.

$$\Rightarrow n \left(P_A^{-1/2} P_X P_B^{-1/2} \right) \leq 1$$

(or)

$$n \left(P_A^{-1/2} P_X P_A^{-1/2} P_B^{1/2} P_A^{-1/2} P_B^{-1/2} \right) \leq 1$$

(or)

$$\left(P_A^{-1/2} P_X P_A^{-1/2} \right)^2 \leq P_A^{-1/2} P_B P_A^{-1/2}$$

Therefore, by operator monotony of $t \rightarrow t^{1/2}$.

$$P_A^{-1/2} P_X P_A^{-1/2} \leq \left(P_A^{-1/2} P_B P_A^{-1/2} \right)^{1/2}$$

$$\text{Hence } P_X \leq P_A^{1/2} \left(P_A^{-1/2} P_B P_A^{-1/2} \right)^{1/2} P_A^{1/2} = P_A \# P_B$$

Next we have check that,

$$\begin{pmatrix} P_A & P_A \# P_B \\ P_A \# P_B & P_B \end{pmatrix} \geq 0$$

$$\Rightarrow n \left(P_A^{-1/2} (P_A \# P_B) P_B^{-1/2} \right) = 1 \text{ since } P_A^{-1/2} (P_A \# P_B) P_B^{-1/2} \text{ is unitary.}$$

Theorem 2.10

Let $P_A, P_B \in M_n(F)$ such that $\alpha P_A \geq P_B \geq \beta P_A$ for some $0 \leq \alpha, \beta \leq 1$ and let Φ be a positive linear map then

$$\Phi(P_A) \# \Phi(P_B) \leq \frac{(\alpha/\beta)^{1/4} + (\beta/\alpha)^{1/4}}{2} \Phi(P_A \# P_B).$$

Proof:

Step 1

Suppose that for a vector f ,

$$\Phi(P_A) = (f, P_A f)$$

$$P_Z = \left(P_A^{-1/2} P_B P_A^{-1/2} \right)^{1/2} \quad \text{and} \quad h = P_A^{1/2} f$$

Since $\alpha P_A \geq P_B \geq \beta P_A$ implies

$$\alpha^{1/2} \geq \left(P_A^{-1/2} P_B P_A^{-1/2} \right)^{-1/2} \geq \beta^{1/2}$$

$$\{\langle f, P_A f \rangle, \langle f, P_B f \rangle\}^{1/2} \leq \frac{(\alpha/\beta)^{1/4} + (\beta/\alpha)^{1/4}}{2} (f, P_A \# P_B f).$$

Step 2

Let h be any vector. Then by Fact 1

$$\langle h, \Phi(P_A) \# \Phi(P_B) h \rangle \leq \langle h, \Phi(P_A) h \rangle^{1/2} \langle h, \Phi(P_B) h \rangle^{1/2}$$

$$= \Psi(P_A)^{1/2} \Psi(P_B)^{1/2}$$

Where Ψ is defined by $\Psi(P_X) = \langle h, \Psi(P_X) h \rangle$

By fact 2: $\Psi(P_X) = T_r P_Y P_X = \langle P_X^{1/2}, \Pi(P_X) P_Y^{1/2} \rangle$ for some $P_Y \in M_n(F)$

Since $\alpha P_A \geq P_B \geq \beta P_A$ implies $\alpha \Pi(P_A) \geq \Pi(P_B) \geq \beta \Pi(P_A)$

$$\Psi(P_A)^{1/2} \Psi(P_B)^{1/2} \leq \frac{(\alpha/\beta)^{1/4} + (\beta/\alpha)^{1/4}}{2} \Psi(P_A \# P_B).$$

Combining with the previous inequality, we get

$$\langle h, \Phi(P_A) \# \Phi(P_B) h \rangle \leq \frac{(\alpha/\beta)^{1/4} + (\beta/\alpha)^{1/4}}{2} \langle h, \Phi(P_A) \# \Phi(P_B) h \rangle$$

Theorem 2.11

Let P_Z in $M_n(F)$ with extremal eigen values a, b for all vectors h in $M_n(F)$.

$$n(h) n(P_Z h) \leq \frac{(\alpha/b)^{1/2} + (b/\alpha)^{1/2}}{2} \langle h, P_Z h \rangle.$$

Proof:

Fact 1: For P_A, P_B in $M_n(F)$ all vectors h and all positive linear maps Φ .

$$\langle h, \Phi(P_A \# P_B) h \rangle \leq \langle h, \Phi(P_A) h \rangle^{1/2} \langle h, \Phi(P_B) h \rangle^{1/2} \quad \text{-----(2)}$$

$$\langle h, P_A \# P_B h \rangle \leq \langle h, P_A h \rangle^{1/2} \langle h, P_B h \rangle^{1/2}$$

Inequality (2),

$$\Phi(P_A \# P_B) \leq \Phi(P_A) \# \Phi(P_B)$$

Where $\Phi(P_A) = P_Z^* P_A P_Z$ for some $n \times k$ fuzzy matrix P_Z .

$$\langle h, P_A \# P_B h \rangle = \langle P_A^{1/2} h, (P_A^{-1/2} P_B P_A^{-1/2}) P_A^{1/2} h \rangle$$

$$\leq n \left(P_A^{1/2} h \right) n \left(\left(P_A^{-1/2} P_B P_A^{-1/2} \right) P_A^{1/2} h \right)$$

$$\langle h, P_A \# P_B h \rangle \leq \langle h, P_A h \rangle^{1/2} \langle h, P_B h \rangle^{1/2}$$

Fact 2: let Φ be a positive fuzzy linear function on $P^*(M_n(F))$. Then there exists P_X in $M_n(F)$ such that $\Phi(P_A) = T_r P_A P_X$.

Hence if $\Pi(P_A): P^*(M_n(F)) \times P^*(M_n(F)) \rightarrow [0,1]$ is the left multiplication by P_A .

$$\Phi(P_A) = \langle h, \Pi(P_A) h \rangle$$

Where the fuzzy inner product is the canonical inner product on $M_n(F)$ and $h = P_X^{1/2}$.

Theorem 2.12

Let P_A, P_B in $M_n(F)$ such that $\alpha P_B(u) \supset P_A(u) \supset \beta P_B(u)$ for some $\alpha, \beta \in [0,1]$. Then $\sum \lambda_j(P_A) \lambda_j(P_B) \leq \frac{\alpha+\beta}{2\sqrt{\alpha\beta}} T_r P_A P_B$.

Proof:

We may assume P_A, P_B in $M_n(F)$ and the inclusion conditions on $P_A(u), P_B(u)$ of $P^*(M_n(F))$.

$$\Rightarrow \alpha \geq |P_B^{-1} P_A| \geq \beta \text{ equivalently } \alpha \geq |P_A P_B^{-1}| \geq \beta$$

$$\therefore n(P_A h) n(P_B h) \leq \frac{\alpha + \beta}{2\sqrt{\alpha\beta}} \langle h, P_A^2 \# P_B^2 h \rangle \dots \dots (3)$$

That is, $\Phi(P_X) = \langle h, P_X h \rangle$, denoted by $\|\cdot\|_1$ or n_1 the trace norm. There exists a unitary V such that $\sum \lambda_j(P_A) \lambda_j(P_B) = n_1(P_A \vee P_B)$

Where $V = \sum v_i h_i \otimes h_i$, Hence making use of the triangle inequality for the trace norm.

$$\sum \lambda_j(P_A) \lambda_j(P_B) \leq \sum n(P_A h_i) n(P_B h_i)$$

Combining with (2), we get

$$\sum \lambda_j(P_A) \lambda_j(P_B) \leq \frac{\alpha + \beta}{2\sqrt{\alpha\beta}} T_r P_A^2 \# P_B^2$$

Next to check that,

$$T_r P_A^2 \# P_B^2 \leq T_r P_A P_B.$$

$$\begin{aligned} \text{there is a unitary } U \text{ such that } P_A^2 \# P_B^2 &= P_A \cup P_B \\ &\Rightarrow T_r P_A^2 \# P_B^2 = T_r P_A \cup P_B \\ &= T_r \left(P_A^{1/2} \cup P_B^{1/2} \right) (P_B^{1/2} P_A^{1/2}) \\ &\leq (T_r P_A \cup P_B \cup^*)^{1/2} (T_r P_A P_B)^{1/2} \\ &\leq (T_r P_A \cup P_B)^{1/2} (T_r P_A P_B)^{1/2} \\ &T_r P_A \cup P_B \leq T_r P_A P_B \end{aligned}$$

III. Conclusion

In this paper, the inequality of fuzzy matrices and its Some Cauchy Schwarz inequality in $M_n(F)$ are discussed. Numerical examples are given to clarify the developed theory and the proposed inequality of fuzzy matrices.

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