

FUZZY PROXIMITY MATRIX
EQUATIONS OF VARIED ORDER

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We advance a concept about fuzzy proximity matrix equations of varied order, and give the family of all its solutions.

Keywords: Fuzzy proximity matrix, Matrix equation of varied order, Greatest solution, Lower solution.

1. Definitions

Definition 1. The matrix

$$R = \begin{bmatrix} r_{11} & \cdots & r_{1n} \\ \vdots & & \vdots \\ r_{n1} & \cdots & r_{nn} \end{bmatrix}$$

is a fuzzy proximity matrix of nth order, if $r_{ii}=1$, $r_{ij}=r_{ji}$, and $r_{ij} \in [0,1]$. The family of all these matrices we denote by \mathcal{R} .

Definition 2 Let $X \in \mathcal{R}$ be unknown and $R, S \in \mathcal{R}$ be given. Write $\hat{K} = \{K, K+1, \dots\}$, where K is a natural number.

The equation

$$X^{\hat{K}} \circ R = S \quad (1)$$

is called the fuzzy proximity matrix equation of varied

order, and means for every $m \in \dot{K}$ (i.e. $m \geq k$),

$$X^m \circ R = \underbrace{X \circ \dots \circ X \circ R}_m = S.$$

where \circ denotes the max-min composition. (fuzzy matrix product).

Definition 3 A is called a \dot{k} type matrix, if $A^{\dot{n}} = A^{\dot{n}+1}$.
 Let \mathcal{I} denote the family of all 1 type idempotent matrices. A type \dot{n} matrix is convergent of index \dot{n} .

3. Propositions

Proposition 1 Let $A \in \mathcal{R}$, then

$$A \subseteq A^2 \subseteq \dots \subseteq A^{\dot{n}-1} = A^{\dot{n}} = \dots \quad (2)$$

where the $A \subseteq C$ denotes $a_{ij} \leq c_{ij}$.

Proposition 2 Let

$$\mathcal{X}(y, \dot{k}) = \{ X \in \mathcal{R} \mid X^{\dot{k}} = y, \dot{K} = \{ k, k+1, \dots \} \} \quad (3)$$

then

$$\mathcal{X}(y, \dot{k}) \neq \emptyset \text{ iff } y = y^2.$$

proof. If $y = y^2$, then

$$y = y^2 = \dots = y^k = y^{k+1} = \dots$$

therefore $y \in \mathcal{X}(y, \dot{k})$, i.e. $\mathcal{X}(y, \dot{k}) \neq \emptyset$.

conversely, if $\mathcal{X}(y, \dot{k}) \neq \emptyset$, assume $A \in \mathcal{X}(y, \dot{k})$, then

$$A^k = A^{k+1} = \dots = y,$$

and $y^2 = A^{2k}$.

but $A^k = A^{2k}$,

thereupon $y = y^2$

Proposition 3 Let

$$\mathcal{X}(R, S, \dot{K}) = \{ X \in \mathcal{R} \mid X^{\dot{k}} \circ R = S, \dot{K} = \{ k, k+1, \dots \} \} \quad (4)$$

and

$$\mathcal{Y}(R, S) = \{ Y \in \mathcal{A} \mid Y \circ R = S \} \quad (5)$$

then $\mathcal{X}(R, S, \dot{K}) \neq \emptyset$ iff $\mathcal{Y}(R, S) \neq \emptyset$

proof. If $\mathcal{X}(R, S, \dot{K}) \neq \emptyset$, let $A \in \mathcal{X}(R, S, \dot{K})$, then $A^{\dot{k}} \circ R = S$.

but, $A^k = A^{k+1} = \dots = A^{2k} = \dots$

This shows $A^k = (A^k)^2$.

Therefore $A^k \in \mathcal{Y}(R, S)$, i.e. $\mathcal{Y}(R, S) \neq \emptyset$.

Conversely, if $\mathcal{Y}(R, S) \neq \emptyset$, then there is a B

$B \in \mathcal{Y}(R, S)$, So that

$$B \circ R = S,$$

and $B = B^2$

By $B = B^2$, we obtained that

$$B = B^2 = \dots = B^k = B^{k+1} = \dots$$

According to $B \circ R = S$, we have

$$B^{\dot{k}} \circ R = S$$

Thereupon $B \in \mathcal{X}(R, S, \dot{K})$, i.e.

$$\mathcal{X}(R, S, \dot{K}) \neq \emptyset.$$

Proposition 4 Let

$$\mathcal{X}(y, m) = \{x \in \mathcal{R} \mid x^m = y, y \in \mathcal{A}\} \quad (6)$$

and

$$\mathcal{X}(y, k) = \{x \in \mathcal{R} \mid x^k = y, y \in \mathcal{A}\} \quad (7)$$

then for $m \geq k$, we have

$$\mathcal{X}(y, m) \supseteq \mathcal{X}(y, k) \quad (8)$$

Proof. For any $X_0 \in \mathcal{X}(y, k)$, then

$$X_0^k = y$$

and

$$y = y^2 \quad (9)$$

Because $m \geq k$, there exists a positive integer α , such that $\alpha k \geq m$.

and then, by (2) and (10), we obtain that

$$X_0^k \subseteq X_0^m \subseteq X_0^{\alpha k},$$

and

$$X_0^{\alpha k} = y^{\alpha} = y = X_0^k$$

Therefore

$$X_0^m = X_0^k = y, \quad \text{i.e.}$$

$$X_0 \in X(y, m).$$

Proposition 5

$$\mathcal{X}(R, S, \dot{K}) = \bigcup_{y \in \mathcal{Y}(S, S)} \mathcal{X}(y, k) \quad (10)$$

where K be not the universal matrix and $K > 1$.

Proof. We can easily obtain that

$$\mathcal{X}(R, S, \dot{K}) = \bigcap_{m \geq k} \left(\bigcup_{y \in \mathcal{Y}(R, S)} \mathcal{X}(y, m) \right)$$

By (9), obtain that

$$\mathcal{X}(y, m) \supseteq \mathcal{X}(y, k),$$

for $m \geq k$.

Therefore

$$\bigcup_{y \in \mathcal{Y}(R, S)} \mathcal{X}(y, m) \supseteq \bigcup_{y \in \mathcal{Y}(R, S)} \mathcal{X}(y, k)$$

for $m \geq k$.

Thereupon

$$\bigcap_{m \geq k} \left(\bigcup_{y \in \mathcal{Y}(R, S)} \mathcal{X}(y, m) \right) = \bigcup_{y \in \mathcal{Y}(R, S)} \mathcal{X}(y, k)$$

i.e.

$$\mathcal{X}(R, S, \dot{K}) = \bigcup_{y \in \mathcal{Y}(R, S)} \mathcal{X}(y, k).$$

In particular

$$\mathcal{X}(y, \dot{k}) = \bigcup_{y \in \mathcal{A}} \mathcal{X}(y, k). \quad (11)$$

Proposition 6 Let $A = (a_{ij}) \in \mathcal{R}$,

$$B = (b_{ij}) \in \mathcal{R} \quad \text{and} \quad A \circ B = C = (c_{ij}), \quad \text{then}$$

$$C \supseteq A \quad (12)$$

and

$$C \supseteq B \quad (13)$$

Proof. In fact, for any $i, j = 1, \dots, n$, we have

$$\begin{aligned} c_{ij} &= \max_{1 \leq k \leq n} \min(a_{ik}, b_{ki}) \\ &= \max \{ \min(a_{ij}, b_{jj}), \min(a_{ii}, b_{ij}), \\ &\quad \max_{k \neq i, j} \min(a_{ik}, b_{kj}) \} \\ &= \max \{ a_{ij}, b_{ij}, \max_{k \neq i, j} \min(a_{ik}, b_{kj}) \} \\ &\geq \max(a_{ij}, b_{ij}). \end{aligned}$$

Therefore $C \supseteq A$ and $C \supseteq B$

Proposition 7 If $R=S$ or $R \leq S$, $S=S^2$, then

$$\mathcal{Y}(R, S) \neq \emptyset$$

Proof. First suppose that $R=S$, then $I \circ R = S$ and

$$I^2 = I = \begin{bmatrix} 1 & & & 0 \\ & 1 & & \\ 0 & & \cdot & \\ & & & \cdot & \\ & & & & 1 \end{bmatrix}$$

Therefore $I \in \mathcal{Y}(R, S)$.

If $R \leq S$ and $S=S^2$, then

$$SoR \subseteq SoS = S.$$

By (12), obtain that

$$SoR \supseteq S.$$

Therefore

$$SoR = S \quad \text{and} \quad S = S^2,$$

i.e. $S \in \mathcal{Y}(R, S).$

3. Examples

Example 1 Let us consider equation (1) with

$$R = \begin{bmatrix} 1 & 0.6 & 0.3 & 0 \\ 0.6 & 1 & 0.1 & 0.2 \\ 0.3 & 0.1 & 1 & 0.7 \\ 0 & 0.2 & 0.7 & 1 \end{bmatrix},$$

$$S = \begin{bmatrix} 1 & 0.6 & 0.4 & 0.4 \\ 0.6 & 1 & 0.4 & 0.4 \\ 0.4 & 0.4 & 1 & 0.7 \\ 0.4 & 0.4 & 0.7 & 1 \end{bmatrix}$$

$$K = \{2, 3, 4, \dots\}$$

According to the paper (2) we obtain that

$$\begin{bmatrix} 1 & 0 & 0 & 0.4 \\ 0 & 1 & 0.4 & 0 \\ 0 & 0.4 & 1 & 0 \\ 0.4 & 0 & 0 & 1 \end{bmatrix} \subseteq X^2 \subseteq \begin{bmatrix} 1 & 0.6 & 0.4 & 0.4 \\ 0.6 & 1 & 0.4 & 0.4 \\ 0.4 & 0.4 & 1 & 0.7 \\ 0.4 & 0.4 & 0.7 & 1 \end{bmatrix}$$

and

$$\begin{bmatrix} 1 & 0 & 0.4 & 0 \\ 0 & 1 & 0 & 0.4 \\ 0.4 & 0 & 1 & 0 \\ 0 & 0.4 & 0 & 1 \end{bmatrix} \subseteq X^2 \subseteq \begin{bmatrix} 1 & 0.6 & 0.4 & 0.4 \\ 0.6 & 1 & 0.4 & 0.4 \\ 0.4 & 0.4 & 1 & 0.7 \\ 0.4 & 0.4 & 0.7 & 1 \end{bmatrix}$$

we obtain that

$$\begin{bmatrix} 1 & 0 & 0 & 0.4 \\ 0 & 1 & 0.4 & 0 \\ 0 & 0.4 & 1 & 0 \\ 0.4 & 0 & 0 & 1 \end{bmatrix}^2 = \begin{bmatrix} 1 & 0 & 0 & 0.4 \\ 0 & 1 & 0.4 & 0 \\ 0 & 0.4 & 1 & 0 \\ 0.4 & 0 & 0 & 1 \end{bmatrix},$$

and so

$$\begin{bmatrix} 1 & 0 & 0.4 & 0 \\ 0 & 1 & 0 & 0.4 \\ 0.4 & 0 & 1 & 0 \\ 0 & 0.4 & 0 & 1 \end{bmatrix}^2 = \begin{bmatrix} 1 & 0 & 0.4 & 0 \\ 0 & 1 & 0 & 0.4 \\ 0.4 & 0 & 1 & 0 \\ 0 & 0.4 & 0 & 1 \end{bmatrix}$$

and

$$\begin{bmatrix} 1 & 0.6 & 0.4 & 0.4 \\ 0.6 & 1 & 0.4 & 0.4 \\ 0.4 & 0.4 & 1 & 0.7 \\ 0.4 & 0.4 & 0.7 & 1 \end{bmatrix}^2 = \begin{bmatrix} 1 & 0.6 & 0.4 & 0.4 \\ 0.6 & 1 & 0.4 & 0.4 \\ 0.4 & 0.4 & 1 & 0.7 \\ 0.4 & 0.4 & 0.7 & 1 \end{bmatrix}$$

According to the paper (3), the solutions of the equations

$$X^2 = \begin{bmatrix} 1 & 0 & 0 & 0.4 \\ 0 & 1 & 0.4 & 0 \\ 0 & 0.4 & 1 & 0 \\ 0.4 & 0 & 0 & 1 \end{bmatrix}$$

and

$$X^2 = \begin{bmatrix} 1 & 0 & 0.4 & 0 \\ 0 & 1 & 0 & 0.4 \\ 0.4 & 0 & 1 & 0 \\ 0 & 0.4 & 0 & 1 \end{bmatrix}$$

are respectively

$$X = \begin{bmatrix} 1 & 0 & 0 & 0.4 \\ 0 & 1 & 0.4 & 0 \\ 0 & 0.4 & 1 & 0 \\ 0.4 & 0 & 0 & 1 \end{bmatrix}$$

and

$$X = \begin{bmatrix} 1 & 0 & 0.4 & 0 \\ 0 & 1 & 0 & 0.4 \\ 0.4 & 0 & 1 & 0 \\ 0 & 0.4 & 0 & 1 \end{bmatrix}$$

The greatest solution of the equation

$$X^2 = \begin{bmatrix} 1 & 0.6 & 0.4 & 0.4 \\ 0.6 & 1 & 0.4 & 0.4 \\ 0.4 & 0.4 & 1 & 0.7 \\ 0.4 & 0.4 & 0.7 & 1 \end{bmatrix}$$

is

$$X = \begin{bmatrix} 1 & 0.6 & 0.4 & 0.4 \\ 0.6 & 1 & 0.4 & 0.4 \\ 0.4 & 0.4 & 1 & 0.7 \\ 0.4 & 0.4 & 0.7 & 1 \end{bmatrix}$$

Therefore, by (10) we obtain

$$\mathcal{C}(R, S, K) = \left\{ X \mid \begin{bmatrix} 1 & 0 & 0 & 0.4 \\ 0 & 1 & 0.4 & 0 \\ 0 & 0.4 & 1 & 0 \\ 0.4 & 0 & 0 & 1 \end{bmatrix} \subseteq X \subseteq \begin{bmatrix} 1 & 0.6 & 0.4 & 0.4 \\ 0.6 & 1 & 0.4 & 0.4 \\ 0.4 & 0.4 & 1 & 0.7 \\ 0.4 & 0.4 & 0.7 & 1 \end{bmatrix} \right\}$$

$$\cup \left\{ X \mid \begin{bmatrix} 1 & 0 & 0.4 & 0 \\ 0 & 1 & 0 & 0.4 \\ 0.4 & 0 & 1 & 0 \\ 0 & 0.4 & 0 & 1 \end{bmatrix} \subseteq X \subseteq \begin{bmatrix} 1 & 0.6 & 0.4 & 0.4 \\ 0.6 & 1 & 0.4 & 0.4 \\ 0.4 & 0.4 & 1 & 0.7 \\ 0.4 & 0.4 & 0.7 & 1 \end{bmatrix} \right\}$$

Example 2 Let us consider equation

$$X^4 = \begin{bmatrix} 1 & 0.4 & 0.8 & 0.5 & 0.5 \\ 0.4 & 1 & 0.4 & 0.4 & 0.4 \\ 0.8 & 0.4 & 1 & 0.5 & 0.5 \\ 0.5 & 0.4 & 0.5 & 1 & 0.6 \\ 0.5 & 0.4 & 0.5 & 0.6 & 1 \end{bmatrix}$$

According to the paper (3) we obtain that the greatest solution of the equation

$$X^4 = \begin{bmatrix} 1 & 0.4 & 0.8 & 0.5 & 0.5 \\ 0.4 & 1 & 0.4 & 0.4 & 0.4 \\ 0.8 & 0.4 & 1 & 0.5 & 0.5 \\ 0.5 & 0.4 & 0.5 & 1 & 0.6 \\ 0.5 & 0.4 & 0.5 & 0.6 & 1 \end{bmatrix}$$

$$\begin{bmatrix} 1 & 0.4 & 0.8 & 0.5 & 0.5 \\ 0.4 & 1 & 0.4 & 0.4 & 0.4 \\ 0.8 & 0.4 & 1 & 0.5 & 0.5 \\ 0.5 & 0.4 & 0.5 & 1 & 0.6 \\ 0.5 & 0.4 & 0.5 & 0.6 & 1 \end{bmatrix}$$

and the lower solutions of this equation be of form

$$X_{ij}(y, \kappa) = \begin{bmatrix} 1 & a_1 & 0.8 & b_1 & b_2 \\ a_1 & 1 & a_2 & a_3 & a_4 \\ 0.8 & a_2 & 1 & b_3 & b_4 \\ b_1 & a_3 & b_3 & 1 & 0.6 \\ b_2 & a_4 & b_4 & 0.6 & 1 \end{bmatrix}$$

Where $a_i=0$, $b_j=0$ and $a_m=0.4$, $b_n=0.5$, for $i, j \in \{1, 2, 3, 4\}$, $i \neq m$, $j \neq n$.

By (12) we obtain

$$S(X, \frac{1}{2}) = \bigcup_{1 \leq i, j \leq 4} \{X \mid \begin{bmatrix} 1 & a_1 & 0.8 & b_1 & b_2 \\ a_1 & 1 & a_2 & a_3 & a_4 \\ 0.8 & a_2 & 1 & b_3 & b_4 \\ b_1 & a_3 & b_3 & 1 & 0.6 \\ b_2 & a_4 & b_4 & 0.6 & 1 \end{bmatrix} \subseteq X$$

$$\subseteq \left\{ \begin{bmatrix} 1 & 0.4 & 0.8 & 0.5 & 0.5 \\ 0.4 & 1 & 0.4 & 0.4 & 0.4 \\ 0.8 & 0.4 & 1 & 0.5 & 0.5 \\ 0.5 & 0.4 & 0.5 & 1 & 0.6 \\ 0.5 & 0.4 & 0.5 & 0.6 & 1 \end{bmatrix} \right\}$$

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