

## *International Journal of Scientific Research and Reviews*

### **A Note on Largest Eigen Fuzzy Set**

**B.Praba<sup>1\*</sup>, M. Logeshwari<sup>2</sup> and R. Mohan<sup>3</sup>**

<sup>1,2</sup>Department of Mathematics,SSN College of Engineering,Chennai - 603110, India

<sup>3</sup> Department of Mathematics, RKM Vivekananda College, Chennai- 600004, India

#### **ABSTRACT**

The concept of finite discrete time Fuzzy Possibilistic Markov Chain (FPMC) on the fuzzy possibility space  $(X, R, \tilde{\Pi})$  is introduced. We analyze the classification of its states. Finally, we give the necessary conditions for the occurrence of the ergodicity and find its steady state using Eigen fuzzy sets.

**KEYWORDS:** Possibility measure, fuzzy possibility, ergodicity, steady state

#### **\*Corresponding author**

**Dr. B. Praba**

Department of Mathematics,  
SSN College of Engineering,  
Chennai - 603110, India.

Email: [prabab@ssn.edu.in](mailto:prabab@ssn.edu.in), Mob No – 9444349085

## 1 INTRODUCTION:

In some circumstances, where the existing information about the system is not enough, the estimation of the probability values is difficult. To overcome these difficulties, the possibility measures have been used since 1965. A Fuzzy Markov Chain<sup>1</sup> was defined with crisp transition probabilities by Bhattacharya in<sup>2</sup> and with fuzzy transition probabilities by Kruse et al<sup>3</sup>. In<sup>4</sup>, the authors defined the fuzzy finite Markov chains based on possibility theory and compared the results of classical Markov chains and FMC. Possibilistic Markov processes and possibilistic Markov chain were defined and analyzed in<sup>6</sup>. Several authors's contributed considerable work in this direction<sup>8, 9, 10, 11, 17</sup>. In<sup>2</sup> Avrachenkov and Sanchez, pointed out the difference between the classical and fuzzy Markov chains.

## 2 PRELIMINARIES:

### DEFINITION 2.1

For every possibility measure  $\Pi$  on  $(X, R)$ , there exists a unique  $R$ -measurable mapping  $\pi : X \rightarrow L$  such that for any  $B \in R$ ,  $\Pi(B) = \sup_{x \in B} \pi(x)$  where  $\pi(x) = \Pi([x]_R)$ ,  $x \in X$ .

### DEFINITION 2.2

Let  $(\Omega, R_\Omega, \Pi_\Omega)$  be another possibility space. A mapping  $f: \Omega \rightarrow X$  is called possibility variable iff it is a  $R_\Omega - R_X$  measurable. Let  $B \in R$ , then

$$\left. \begin{aligned} \Pi(B) &= \Pi_\Omega(f^{-1}(B)) \\ \pi(x) &= \sup_{f(\omega) \in [x]_R} [\Pi_\Omega(\omega)], \forall x \in X \end{aligned} \right\} (2.1)$$

### DEFINITION 2.3

Let  $f: \Omega \rightarrow X$  be a possibility variable on  $(X, R)$ . By the transformation of possibility measure,  $\Pi$  is a possibility measure on  $(X, R)$ .

$$\left. \begin{aligned} \Pi_f(B) &= (\Pi_\Omega^f R)(B), \forall B \in R = \Pi_\Omega(f^{-1}(B)) \\ \Pi_f(x) &= \sup_{f(\omega) \in [x]_R} [\Pi_\Omega(\omega)], \forall x \in X \end{aligned} \right\} (2.2)$$

Where  $\Pi_f$  is called as the possibility measure of  $f$

## 3 PROPERTIES AND CLASSIFICATION OF FPMC

### 3.1 Fuzzy Possibilistic Markov chain ( FPMC)

In this subsection, we have defined some new concepts and have introduced FPMC. We have also generalized the properties of classical Markov chain to FPMC using max-min composition.

**THEOREM 3.1**

Let  $\tilde{A}^\Pi = \{x, \mu_{\tilde{A}^\Pi}(x) = \tilde{B}_x\}$  be a normalized type 2 fuzzy set defined on  $X$  and  $\tilde{B}_x$ .

Then  $\tilde{\Pi}$  forms a possibility space  $(X, R, \tilde{\Pi})$  such that there exist a possibility variable  $f$  with the possibility distribution on  $\mu_{\tilde{A}^\Pi}$ .

**PROOF:**

Let  $\tilde{A}^\Pi$  be as given in the statement and let  $\tilde{\Pi}: R \rightarrow F([0, 1])$  be the fuzzy Possibility such that  $\tilde{\Pi}(B) = \sup_{x \in B} [\mu_{\tilde{A}^\Pi}(x)] = \sup_{x \in B} \tilde{B}_x$  for all  $B \in R$ . Clearly,  $\tilde{\Pi}(\Phi) = 0$ . Since the given type 2 fuzzy set is normalized, there exists an element with membership grade  $(1, 1, 1)$  and it is the supremum among all fuzzy numbers on  $[0, 1]$ . Hence  $\tilde{\Pi}(X) = \sup_{x \in X} [\tilde{B}_x] = (1, 1, 1)$ . To prove that  $\tilde{\Pi}(\cup_i B_i) = \sup_i (\tilde{\Pi}(B_i))$ , where  $\{B_i\}$  is an arbitrary collection of subsets of  $X$ , let  $C = \cup_i B_i$ . Then  $\tilde{\Pi}(C) = \sup_{x \in C} [\tilde{B}_x] = \tilde{B}_{x_1}$ . Obviously, the corresponding  $x_1$  is in at least one of the  $B_i$ 's. Hence,  $\sup_i [\tilde{\Pi}(B_i)] = \sup_i \{ \sup_{x \in B_i} [\tilde{B}_x] \} = \tilde{B}_{x_1}$  and  $\tilde{\Pi}(\cup_i B_i) = \sup_i (\tilde{\Pi}(B_i))$ . This proves that the fuzzy possibility  $\tilde{\Pi}$  is a possibility measure on  $(X, R)$  with possibility distribution  $\tilde{\pi}(x) = \mu_{\tilde{A}^\Pi}(x)$  i.e.,  $(X, R, \tilde{\Pi})$  is possibility space. By (2.2) and (2.3),  $f$  is a possibility distribution.

Hence,  $\tilde{\Pi}f = \mu_{\tilde{A}^\Pi}$ .

**Note:** The possibility space  $(X, R, \tilde{\Pi})$  induced by the fuzzy possibility  $\tilde{\Pi}$  is named as fuzzy possibility space and the possibility distribution is called fuzzy possibility distribution.

**DEFINITION 3.1**

The set of possibility variables  $\{X(t); t \in T\}$ , defined on the fuzzy possibility space  $(X, R, \tilde{\Pi})$  is called the fuzzy possibilistic stochastic process (FPSP).

**DEFINITION 3.2**

The Fuzzy Possibilistic Stochastic Process (FPSP) satisfying the Markovian property is called the fuzzy possibilistic Markov chain (FPMC),

And  $\max_j \{\tilde{\pi}_{i,j}\} = (1, 1, 1)$ . The transition between the states of the FPMC can be viewed as a fuzzy relation with fuzzy relation matrix  $\tilde{H}$ . Hence, instead of matrix multiplication, we are using composition of two fuzzy relations (max-min composition).

**DEFINITION 3.3**

AFPMC is said to be homogeneous, if the fuzzy transition possibility from state 'i' at step m to state 'j' at step n does not depend on m and n, but only on the difference n-m.

$$i.e., \tilde{\pi}_{ij}(n,m) = \tilde{\pi}_{ij}(n-m) \quad \tilde{H}(n,m) = \tilde{H}(n-m).$$

**THEOREM 3.2**

Consider a homogeneous FPMC  $\{X(n), n = 0, 1, \dots\}$  with m states and with fuzzy transition possibility  $\tilde{\pi}_{ij}$ . Then for all  $n_1, n_2 \geq 0, \tilde{H}(n_1+n_2) = \tilde{H}(n_1) \otimes \tilde{H}(n_2)$ .

**PROOF:**

Consider

$$\begin{aligned} \tilde{\Pi}^{n_1+n_2} &= \tilde{\Pi}\{X(n_1 + n_2) = j | X(0) = i\} \\ &= \max_{k=0}^{n_1+n_2} \min[\tilde{\Pi}_{ik}(n_1), \tilde{\Pi}_{kj}(n_2)] \end{aligned}$$

Note that  $\max_i(\tilde{\Pi}_{ij})$  hence proved and it is called the Chapman-Kolmogorov equation of FPMC. Hence by induction  $\tilde{H}(n) = \tilde{H}$ .

**DEFINITION 3.4**

Consider a FPMC with m states. Then the fuzzy possibility of being in state I at nth step is given by,  $\tilde{p}_i(n) = \tilde{\Pi}\{X(n) = i\}$ . And the row vector,  $\tilde{P}(n) = [ \tilde{p}_0(n), \tilde{p}_1(n), \dots, \tilde{p}_m(n) ]$  is called the state fuzzy possibility distribution at the nth step. And we denote the limiting state fuzzy possibility distribution as  $\lim_{n \rightarrow \infty} \tilde{P}(n) = \tilde{\Gamma}$ . □

**3.2 Classification Of The States Of Afpmc**

In this subsection, we are considering a homogeneous FPMC with m states.

**DEFINITION 3.5**

Let  $\tilde{F}_{ij} = \sup_i(\tilde{f}_{ij}(n))$  where  $\tilde{f}_{ij}(n)$  the fuzzy possibility of the first time visit to stage j

In the probability space, positive recurrent is defined in terms of mean recurrence time where it is not possible for FPMC. If we define the mean recurrence time for a recurrent state 'i' as  $\beta_i =$

$\sup_k[\inf(k, \tilde{f}_{ii}(k))]$ , since the k-values are positive integers and  $\tilde{f}_{ii}(k)$  are triangular fuzzy numbers

on [0,1], we get  $[\inf(k, \tilde{f}_{ii}(k))] = \tilde{f}_{ii}(k)$  provided  $k \neq 0$ . Hence,  $\beta_i = [\sup_k \tilde{f}_{ii}(k)] = \tilde{f}_{ii} = (1, 1, 1)$  which is always

finite. Hence the positive recurrence for a state of FPMC is defined as follows.

**DEFINITION 3.6**

A state 'i' is transient iff there is a fuzzy possibility that the process will never return to state 'i' and  $\tilde{F}_{ii} < (1,1,1)$ .

**DEFINITION 3.7**

If the powers of the FTPM  $\tilde{H}$  converge in n steps to a non-periodic solution, then the associated FPMC is said to be a periodic.

**4 ERGODICITY OF FPMC**

For FPMC, the limiting FTPM will have stationary solutions depends on the initial state. Hence, in <sup>1</sup> the ergodicity is defined as follows.

**DEFINITION 4.1**

A FPMC is said to be ergodic if it is aperiodic and has limiting FTPM with identical rows.

**THEOREM 4.1**

Let  $\tilde{H}_{3 \times 3}$  be the fuzzy transition possibility matrix of a FPMC with three states. Then FPMC is ergodic, if  $\tilde{H}$  possesses the following properties.

- $\tilde{H}$  has at least one column  $j_1$  in such a way that all the entries of  $j_1$  are equal to  $(1,1,1)$ .
- If  $\max_i(\tilde{h}_{ij})$  for each  $j$  (excluding the  $j_1^{th}$  column and row) are not the diagonal entries, then  $\tilde{h}_{j_1 k} \geq \min \max_i(\tilde{h}_{ij}) \forall j$  and for some  $k$ . If there is  $\tilde{h}_{k_1 k_1} \geq \min \max_i(\tilde{h}_{ij})$  a diagonal entry, then  $\tilde{h}_{j_1 k} \geq \tilde{h}_{k_1 k_1}$ .
- If a diagonal entry  $\tilde{H}_{ii} (i \neq j_1)$  of  $\tilde{H}$  is the largest element in  $i^{th}$  column, then  $\tilde{h}_{j_1 i} \geq \tilde{h}_{ij}$ . If there is  $\tilde{h}_{jj} \geq \tilde{h}_{ij}$  then  $\tilde{h}_{j_1 i} \geq \tilde{h}_{jj}$ .

**PROOF:**

To get the rows of the limiting FTPM to be identical, the FTPM should be converge and the entries in each column of the limiting. FTPM should be equal. Let us assume that  $\tilde{H}$  satisfies the above three properties. Now we prove that there exist a limiting FTPM with identical rows corresponding to  $\tilde{H}$ .

**CASE: 1**

Since the row of the FPMC are fuzzy possibility distributions, the row maximum of each row of  $\tilde{H}$  is (1,1,1). Let the entries of  $i^{th}$  column be equal to (1,1,1). Then, this column entries will be retained in the higher powers of  $\tilde{H}$ . In the remaining part of this proof, we have considered by  $\max_i(\tilde{h}_{ij})$  for each  $j$  excluding  $j^{th}$  column and row.

**CASE: 2**

Let none of the column maximums be the diagonal entries of  $\tilde{H}$ . Since the given FPMC has three states and the diagonal entries will not be the column maximums, the maximum entries of the  $i^{th}$  and  $j^{th}$  column will be  $\tilde{h}_{ij}$ . Let  $\tilde{h}_{j_1i} \geq \min\{\tilde{h}_{ij}, \tilde{h}_{ji}\}$ . Since each row's  $j_1^{th}$  entry is (1,1,1) and  $\tilde{h}_{j_1i} \geq \min\{\tilde{h}_{ij}, \tilde{h}_{ji}\}$ , the entries of  $i^{th}$  column  $\tilde{H}^2$  are equal to  $\tilde{h}_{j_1i}$ . And in the  $j^{th}$  column  $\tilde{H}^2$ ,  $\tilde{h}_{j_1j}^2$  ( $\because \tilde{h}_{j_1j}^2 = \tilde{h}_{ij}$  if  $\tilde{h}_{j_1i} \geq \tilde{h}_{ij}$  or  $\tilde{h}_{j_1j}^2 = \tilde{h}_{j_1i}$  if  $\tilde{h}_{j_1i} \geq \tilde{h}_{ij}$ ). Even though  $\tilde{h}_{j_1i} \geq \min\{\tilde{h}_{ji}, \tilde{h}_{ij}\}$ , if there is a diagonal entry  $\tilde{h}_{jj} \geq \min\{\tilde{h}_{ij}, \tilde{h}_{ji}\}$ , then  $\tilde{h}_{jj}^2 = \tilde{h}_{jj}$ . If  $\tilde{h}_{j_1i} \geq \tilde{h}_{jj}$  then the entries of the  $i^{th}$  column of  $\tilde{H}^2$  are equal to  $\tilde{h}_{j_1i}$ . Since  $\tilde{h}_{j_1i} \geq \tilde{h}_{jj}$  and  $\tilde{h}_{jj} \geq \min\{\tilde{h}_{ji}, \tilde{h}_{ij}\}$ ,  $\tilde{h}_{j_1j}^2$  [ $\tilde{h}_{j_1j}^2 = \tilde{h}_{ij}$  if  $\tilde{h}_{j_1i} \leq \tilde{h}_{ij}$  or  $\tilde{h}_{j_1j}^2 = \tilde{h}_{j_1i}$  if  $\tilde{h}_{j_1i} \geq \tilde{h}_{ij}$ ] is the  $\tilde{H}_{ki}^2 = \tilde{h}_{j_1i}$ . Hence we get the limiting FTPM with identical rows

**CASE: 3**

Suppose that  $\max_k(\tilde{h}_{kj}) = \tilde{h}_{ij}$  and  $\tilde{h}_{j_1i} \geq \tilde{h}_{ii}$ . Then  $\tilde{h}_{j_1j}^2 = \min\{\tilde{h}_{ii}, \tilde{h}_{ij}\}$ . Since  $\tilde{h}_{j_1i} \geq \tilde{h}_{jj}$  and  $\tilde{h}_{jj} \geq \min\{\tilde{h}_{ii}, \tilde{h}_{ij}\}$ ,  $\tilde{h}_{j_1j}^2$  [ $\tilde{h}_{j_1j}^2 = \tilde{h}_{ij}$  if  $\tilde{h}_{j_1i} \geq \tilde{h}_{ij}$  or  $\tilde{h}_{j_1j}^2 = \tilde{h}_{j_1i}$  if  $\tilde{h}_{j_1i} \leq \tilde{h}_{ij}$ ] will be the largest in the  $j^{th}$  column. In  $\tilde{H}^3$ , all the entries of  $j^{th}$  column become equal to  $\tilde{h}_{j_1j}^2$ . Hence we attain the limiting FTPM with identical rows.

**THEOREM 4.2**

For a homogeneous irreducible ergodic FPMC, let  $\tilde{\Gamma}_j = \lim_{n \rightarrow \infty} \tilde{\pi}_{ij}^n, j \geq 0$ . Then,

$\tilde{\Gamma} = (\tilde{\Gamma}_1, \tilde{\Gamma}_2, \dots, \tilde{\Gamma}_m)$  is the greatest eigen fuzzy set (EFS) of  $\tilde{\Gamma} \otimes \tilde{H} = \tilde{\Gamma}$  such that  $\oplus_i \tilde{\Gamma}_i = (1, 1, 1)$ . And  $\tilde{\Gamma}$  is called the steady state of FPMC.

**PROOF:**

The existence of  $\tilde{\Gamma}_j = \lim_{n \rightarrow \infty} \tilde{\pi}_j^n$

$$\begin{aligned} \tilde{\pi}(X(n + 1) = j) &= \max_i \{ \min [ \tilde{\pi}(X(n + 1) = j) | X(n) = i ] \} \\ &= \max_i \{ \min [ \tilde{\pi}_i(n), \tilde{\pi}_{ij} ] \} \end{aligned}$$

$$\begin{aligned} \lim_{n \rightarrow \infty} [\tilde{\pi}_j(n + 1)] &= \lim_{n \rightarrow \infty} \max_i \{ \min(\tilde{\pi}_i(n), \tilde{\pi}_{ij}) \} \\ &= \max_i \{ \min [ \lim_{n \rightarrow \infty} \tilde{p}_i(n), \tilde{\pi}_{ij} ] \} \\ &= \max_i \{ \min [ \tilde{\Gamma}_i, \tilde{\pi}_{ij} ] \} \end{aligned}$$

$$\tilde{\Gamma}_j = \tilde{\Gamma}_i \otimes \tilde{\pi}_{ij}, j \geq 0$$

$$\tilde{\Gamma} = \tilde{\Gamma} \otimes \tilde{H}, j \geq 0.$$

Since the given FPMC is ergodic, the rows of its limiting transition matrix are equal to the greatest EFS of the fuzzy relation defined by P which has been proved in [1], the row soft limiting  $FTPM \tilde{H}$  are equal to the greatest EFS of  $\tilde{\Gamma} = \tilde{\Gamma} \otimes \tilde{H}$  and also  $\tilde{\Gamma}_j = \lim_{n \rightarrow \infty} \tilde{\pi}_j^n, \tilde{\Gamma} = (\tilde{\Gamma}_1, \tilde{\Gamma}_2, \dots, \tilde{\Gamma}_m)$  is the greatest EFS of  $\tilde{\Gamma} = \tilde{\Gamma} \otimes \tilde{H}$  and  $\oplus_i \tilde{\Gamma}_i = (1, 1, 1)$ .

#### 4 EXAMPLE

The *FTPM* of FPMC is,

$$\tilde{H} = \begin{pmatrix} (0.1, 0.3, 0.4) & (1, 1, 1) & (0.3, 0.4, 0.6) \\ (0.1, 0.2, 0.5) & (1, 1, 1) & (0.3, 0.5, 0.6) \\ (0.1, 0.4, 0.5) & (1, 1, 1) & (0.4, 0.6, 0.7) \end{pmatrix}$$

$$\text{Then, } \tilde{H}^2 = \tilde{H}^3 \dots = \begin{pmatrix} (0.1, 0.4, 0.5) & (1, 1, 1) & (0.3, 0.5, 0.6) \\ (0.1, 0.4, 0.5) & (1, 1, 1) & (0.3, 0.5, 0.6) \\ (0.1, 0.4, 0.5) & (1, 1, 1) & (0.4, 0.6, 0.7) \end{pmatrix}$$

Since the limiting *FTP M* converges at  $T = 2$  with non-identical rows, the given FPMC is not ergodic. And the steady state for this FPMC will not exist

#### 5 CONSUMMATION

In this article, we have considered a map, called fuzzy possibility on the ample field of the universe X and have proved that if there exists a normalized type 2 fuzzy set on X, then the fuzzy possibility constructs a possibility space on X. On this fuzzy possibility space, we have defined the FPMC and the rows of its *FTPM*  $\tilde{H}$ . And we have classified its states. We have considered a FPMC with three states and have found out the necessary conditions for its ergodicity. we have proved that, if a FPMC is ergodic, then its steady state is the greatest EFS of the fuzzy relation defined by  $\tilde{H}$ .

#### ACKNOWLEDGMENT

We thank the management of SSN Institutions and the Principal for providing necessary facilities to carry out this work.

#### REFERENCES

1. Earnest Lazarus J, Piriya Kumar. V and Sreevinotha, On the Differentiability of Fuzzy Transition Probability of Fuzzy Markov Chains. *International Research Journal of Engineering and Technology (IRJET)*. 2015; 2(8). 201-208
2. Avrachekov K. E and Sanche. E. Fuzzy Markov Chain. Specificities and Properties. *Fuzzy Optimization and Decision Making* 2002; 1: 143-159.
3. Kruse R, Emden R. B and Cordes. R. Processor Power Considerations-An Application of Fuzzy Markov Chains. *Fuzzy Sets and Systems* 1987; 21: 289-299.
4. Bhattacharyya M. Fuzzy Markovian Decision Process. *Fuzzy Sets and Systems* 1998; 99:273-282.
5. Klir G. J and Yuan. B. *Fuzzy Sets and Fuzzy Logic. Theory and Applications*. Prentice Hall, 2002:356-361
6. Buckley J. J and Eslami. E. Fuzzy Markov Chains Uncertain Probabilities. *MathWare and Soft Computing* 2002; 9: 33-41.
7. Sujatha. R and Praba. B. A Classification of Fuzzy Markov Model. *Proceeding of the International Conference on Mathematics and Computer Science*. 2007; 494-496.
8. Sujatha. R and Praba. B. Analysis of Fuzzy Markov Model Using Fuzzy Relational Equations. *The Journal on Fuzzy Mathematics* 2008; 16
9. Buckley J. J, Feuring and Hayashi. Y. Fuzzy Markov Chains, *Proc. 9th IFSA World Congress and 20th NAFIPS International Conference*, 2001:2708-2711.
10. Janssen. H, Cooman G. D. and Kerre. E. K First Result for Mathematical Theory of Possibilistic Markov Processes. *Proc. of Information Processing and Management of Uncertainty in Knowledge Based System*. 1996; II : 1425-1431.
11. Praba. B and Sujatha. R Fuzzy Markov Model for Web Testing. *International Conference on Advances in Computing, Control, and Telecommunication Technologies* 2007; 21: 111-120.
12. Qingsong Wang, Ming Cheng, Zhe Chen and Zheng Wang. Steady-State Analysis of Electric Springs With a Novel Control. *IEEE Transactions on Power Electronics*. 2015; 30 (12).
13. Sanchez E. Eigen Fuzzy Sets and Fuzzy Relations. *J. Math. Analysis and Applications* 1981; 81:399-421.
14. Sanchez E. Resolution of Eigen Fuzzy Sets Equations, *Fuzzy Sets and Systems* 1978; 169-74.
15. Thomson. M. G. Convergence of Powers of a Fuzzy Matrix. *J. Math. Analysis and Applications* 1977; 57: 476- 480.

16. Wang.Y Multiscale uncertainty quantification based on a generalized hidden Markov model. Journal of Mechanical Design 2011; 133: 310-321
17. Yoshida.Markov.Y. Chains with a transition possibility measure and fuzzy dynamic programming. Fuzzy Sets and Systems 1994; 66: 39-57.