

# A comparison of some T-norms and T-conorms over the steady state of a fuzzy Markov chain

Juan Carlos Figueroa-García<sup>1</sup>

Universidad Distrital Francisco José de Caldas, Bogotá – Colombia  
jcfigueroag@udistrital.edu.co \*

**Abstract.** This paper shows a comparison of several T-norms and T-conorms used to compute the steady state of a fuzzy Markov chain. The effect of every of the selected norms over the mean and variance of a fuzzy Markov chain is evaluated using some simulations. Some recommendations and concluding remarks are given.

**Keywords:** Fuzzy T-norms, Fuzzy T-conorms, Fuzzy Markov chains

## 1 Motivation

Fuzzy Markov chains is one of the most popular fuzzy stochastic process (see Sanchez [7,8], Avrachenkov & Sanchez [3,2], and Araiza, Xiang, Kosheleva & Skülj [1]) so there are different algorithms, fuzzy relations and compositions to compute the stable state possibilities of the fuzzy process.

Most of published works on fuzzy Markov chains apply the max – min composition to obtain the steady state of the process (see Figueroa-García [4,5]), so there is the possibility of using other compositions to see the effect over its steady state.

It is well known that different T-norms and T-conorms lead to closer/wider results in fuzzy inference systems, so we want to evaluate the effect of several norms such as Gödel, Archimedean, Łukasiewicz, Hamacher, Sugeno–Weber, and Yager T-norms (see Klement et al. [6]) over the mean/variance of a fuzzy Markov chain (a.k.a weak inference of a stochastic process).

## 2 Fuzzy Markov chains

Let  $\mathcal{P}(X)$  be the class of all crisp sets,  $\mathcal{F}(X)$  is the class of all fuzzy sets,  $\mathcal{F}_1(X)$  is the class of all convex fuzzy sets, and  $I = [0, 1]$  be the set of values in the unit interval. A fuzzy set namely  $A$  is characterized by a membership function  $\mu_A : X \rightarrow I$  defined over a universe of discourse  $x \in X$ . Thus, a fuzzy set  $A$  is the set of ordered pairs  $x \in X$  and its membership degree,  $\mu_A(x)$ , i.e.,

$$A = \{(x, \mu_A(x)) \mid x \in X\}. \quad (1)$$

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\* Juan Carlos Figueroa-García is Assistant Professor at the Universidad Distrital Francisco José de Caldas, Bogotá - Colombia

Let us denote  $\mathcal{F}_1(\mathbb{R})$  as the class of all fuzzy numbers. Avrachenkov & Sanchez (see [3]) defined a fuzzy Markov chains as follows.

**Definition 1** Let  $S = \{1, 2, \dots, n\}$ , then a finite fuzzy set or a fuzzy distribution on  $S$  is defined by a map  $x : S \rightarrow [0, 1]$  represented by a vector  $\mathbf{x} = \{x_1, x_2, \dots, x_n\}$  where  $0 \leq x_i \leq 1$ ,  $i \in S$ . The set of all fuzzy sets is denoted by  $\mathcal{F}(S)$ .

A fuzzy relation matrix  $P$  on the cartesian product  $S \times S$  where  $P$  is defined by a matrix  $\{p_{ij}\}_{i,j=1}^n$ , with  $0 \leq p_{ij} \leq 1$ ,  $i, j \in S$ . This fuzzy relationship matrix  $P$  allows us to define the transition matrix of the  $m$  states of the Markov chain between each time instant  $t$  in the following form:

**Definition 2** At each instant  $t$ ,  $t = 1, 2, \dots, n$ , the state of the stochastic process is described by the fuzzy set  $x^{(t)} \in \mathcal{F}(S)$ . The transition law of the Markov chain is given by the fuzzy relation  $P$  as follows, at the instant  $t$ ,  $T = 1, 2, \dots, n$ .

$$\mathbf{x}^{(t+1)} = \bigvee_{i \in S} \{\mathbf{x}^{(t)} \wedge p_{ij}\}, j \in S. \quad (2)$$

where  $i$  and  $j$  are the initial and final state of the transition  $i, j = 1, 2, \dots, m$  and  $\mathbf{x}^{(0)}$  is the initial fuzzy set or the initial distribution of the process.

Like in probabilistic Markov chains, the convergence of the chain in a fuzzy environment can be found using a time limiting criterion, so the powers of a fuzzy transition matrix  $P$  are:

$$p_{ij}^t \triangleq \bigvee_{k \in S} \{p_{ik} \wedge p_{kj}^{t-1}\} \quad (3)$$

or in a matrix form:

$$P^t \triangleq P \circ P^{t-1} \quad (4)$$

where  $\circ$  denotes a fuzzy composition, and the limiting distribution of  $P$  is:

**Definition 3** Let the powers of the fuzzy transition matrix converge in  $\tau$  steps to a non periodic solution, then the associated fuzzy Markov chain is called non periodic (or aperiodic) and  $P^* = P^\tau$  is called a limiting fuzzy transition matrix.

Then the main idea is to evaluate the effect of several T-norms and T-conorms over the expected value and variance of  $P^\tau$ , as defined as follows:

$$E(P) = \frac{\sum_i i \cdot \check{x}_i}{\sum_i \check{x}_i}, \quad (5)$$

$$V(P) = \frac{\sum_i (i - E(P))^2 \cdot \check{x}_i}{n - 1}, \quad (6)$$

where  $i$  is the initial distribution of  $P$  and  $\check{x}_i$  is the  $i_{th}$  element of the limiting distribution of  $P^\tau$ .

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