

# A New Universal Approximation Result For Fuzzy Systems, Which Reflects CNF–DNF Duality

Irina Perfilieva<sup>1</sup> and Vladik Kreinovich<sup>2</sup>

<sup>1</sup> Dept of Natural Science  
University of Ostrava  
70103 Ostrava 1, Czech Republic  
email Irina.Perfilieva@osu.cz

<sup>2</sup>Department of Computer Science  
University of Texas at El Paso  
El Paso, TX 79968, USA  
email vladik@cs.utep.edu

## Abstract

There are two main fuzzy system methodologies for translating expert rules into a logical formula: In Mamdani's methodology, we get a DNF formula (disjunction of conjunctions), and in a methodology which uses logical implications, we get, in effect, a CNF formula (conjunction of disjunctions). For both methodologies, universal approximation results have been proven which produce, for each approximated function  $f(x)$ , two different approximating relations  $R_{\text{DNF}}(x, y)$  and  $R_{\text{CNF}}(x, y)$ . Since in fuzzy logic, there is a known relation  $F_{\text{CNF}}(x) \leq F_{\text{DNF}}(x)$  between CNF and DNF forms of a propositional formula  $F$ , it is reasonable to expect that we would be able to prove the existence of approximations for which a similar relation  $R_{\text{CNF}}(x, y) \leq R_{\text{DNF}}(x, y)$  holds. Such existence is proved in our paper.

# 1 Introduction

## 1.1 Fuzzy control: in brief

Fuzzy control (see, e.g., [9]) is a methodology that translates the expert's if-then rules of the type

$$\text{if } A_i(x) \text{ then } B_i(y), \quad 1 \leq i \leq N, \quad (1)$$

or

$$\text{if } A_{i_1}(x_1) \text{ and } \dots \text{ and } A_{i_n}(x_n) \text{ then } B_i(y), \quad (2)$$

in which the properties  $A_i(x)$  and  $B_j(y)$  are described by using words from natural languages (such as “ $x$  is small”), into a *control strategy*, i.e., into a function  $f : X \rightarrow Y$  describing what exactly control we should apply for a given input  $x \in X$ . This methodology consists of three major steps:

- first, we *formalize* each “linguistic” property  $A_i(x)$  or  $B_i(y)$  as a *fuzzy set*, i.e., as a function  $A_i : X \rightarrow [0, 1]$  which describes, for each object  $x \in X$ , to what extent this property holds for this  $x$  (e.g., to what extent  $x$  is small);
- then, we *combine* these fuzzy sets into a *fuzzy relation*, i.e. a function  $R(x, y) : X \times Y \rightarrow [0, 1]$  which describes, for each input  $x \in X$  and for each possible output  $y \in Y$ , to what extent this particular outputs satisfies the expert's rules;
- finally, we apply some *defuzzification procedure* to the *fuzzy relation*  $R(x, y)$ , and get the desired control strategy, as a function  $\tilde{f} : X \rightarrow Y$ .

## 1.2 Mamdani's (DNF) approach

In most practical application of fuzzy control, *Mamdani's* approach is used in the combination (second) step. In this approach, the fuzzy relation  $R(x, y)$  is represented by a logical formula

$$(A_1(x) \& B_1(y)) \vee \dots \vee (A_N(x) \& B_N(y)), \quad (3)$$

or as

$$(A_{11}(x) \& \dots \& A_{1n}(x_n) \& B_1(y)) \vee \dots \vee (A_{N1}(x_1) \& \dots \& A_{Nn}(x_n) \& B_N(y)) \quad (4)$$

where ‘&’ and ‘ $\vee$ ’ stand for connectives of conjunction and disjunction respectively. In logical terms, we have a disjunction of conjunctions  $A_i(x) \& B_i(y)$ , i.e., a formula in a Disjunctive Normal Form – DNF.

Then, we select an interpretation: a t-norm  $f_{\&} : [0, 1] \times [0, 1] \rightarrow [0, 1]$  for conjunction and a t-conorm  $f_{\vee} : [0, 1] \times [0, 1] \rightarrow [0, 1]$  for disjunction (see, e.g.,

[4, 10]), and use these operations in the corresponding formulas (3) and (4), resulting in:

$$R_{\text{DNF}}(x, y) = f_{\vee}(f_{\&}(A_1(x), B_1(y)), \dots, f_{\&}(A_N(x), B_N(y))), \quad (5)$$

$$R_{\text{DNF}}(x, y) = f_{\vee}[f_{\&}(A_{11}(x), \dots, A_{1n}(x_n), B_1(y)), \dots, f_{\&}(A_{N1}(x_1), \dots, A_{Nn}(x_n), B_N(y))]. \quad (6)$$

### 1.3 Logical implication (CNF) approach

From the logical viewpoint, it is somewhat more natural to represent the fuzzy relation  $R(x, y)$  as a conjunction of implications:

$$(A_1(x) \rightarrow B_1(y)) \& \dots \& (A_N(x) \rightarrow B_N(y)), \quad (7)$$

or

$$((A_{11}(x) \& \dots \& A_{1n}(x_n)) \rightarrow B_1(y)) \& \dots \& ((A_{N1}(x_1) \& \dots \& A_{Nn}(x_n)) \rightarrow B_N(y)). \quad (8)$$

In this case, we select the interpretation:  $f_{\&} : [0, 1] \times [0, 1] \rightarrow [0, 1]$  for conjunction and  $f_{\rightarrow} : [0, 1] \times [0, 1] \rightarrow [0, 1]$ , for implication and use these operations in the corresponding formulas (7) and (8), resulting in:

$$R_{\rightarrow}(x, y) = f_{\&}(f_{\rightarrow}(A_1(x), B_1(y)), \dots, f_{\rightarrow}(A_N(x), B_N(y))), \quad (9)$$

$$R_{\rightarrow}(x, y) = f_{\&}[f_{\rightarrow}(f_{\&}(A_{11}(x), \dots, A_{1n}(x_n)), B_1(y)), \dots, f_{\rightarrow}(f_{\&}(A_{N1}(x_1), \dots, A_{Nn}(x_n)), B_N(y))]. \quad (10)$$

In particular, since in classical logic  $A \rightarrow B$  is equivalent to  $\neg A \vee B$ , and  $(A_1(x) \& \dots \& A_n) \rightarrow B$  to  $\neg A_1 \vee \dots \vee \neg A_n \vee B$ , it makes sense to consider representations of formulas (7) and (8) in the following form

$$(\neg A_1(x) \vee B_1(y)) \& \dots \& (\neg A_N(x) \vee B_N(y)), \quad (11)$$

or

$$(\neg A_{11}(x) \vee \dots \vee \neg A_{1n}(x_n) \vee B_1(y)) \& \dots \& (\neg A_{N1}(x_1) \vee \dots \vee \neg A_{Nn}(x_n) \vee B_N(y)). \quad (12)$$

In logical terms, we have a conjunction of disjunctions  $A_i(x) \& B_i(y)$ , i.e., a formula in a Conjunctive Normal Form – CNF.

In DNF, we have outside disjunction and inside conjunctions; in CNF, the roles of disjunction and conjunction are reversed: we have outside conjunction and inside disjunctions. In logic, conjunction and disjunction are often called *dual* logical operations; in view of this terminology, CNF and DNF are also often called *dual* forms.

Then, we select the interpretation: a t-norm  $f_{\&} : [0, 1] \times [0, 1] \rightarrow [0, 1]$  for conjunction, a t-conorm  $f_{\vee} : [0, 1] \times [0, 1] \rightarrow [0, 1]$  for disjunction and a *fuzzy negation*  $f_{\neg} : [0, 1] \rightarrow [0, 1]$ , and use these operations in the corresponding formulas (11) and (12), resulting in:

$$R_{\text{CNF}}(x, y) = f_{\&}(f_{\vee}(f_{\neg}(A_1(x)), B_1(y)), \dots, f_{\vee}(f_{\neg}(A_N(x)), B_N(y))), \quad (13)$$

$$R_{\text{CNF}}(x, y) = f_{\&}[f_{\vee}(f_{\neg}(A_{11}(x)), \dots, f_{\neg}(A_{1n}(x_n)), B_1(y)), \dots, f_{\vee}(f_{\neg}(A_{N1}(x_1)), \dots, f_{\neg}(A_{Nn}(x_n)), B_N(y))]. \quad (14)$$

#### 1.4 Relation between DNF and CNF approaches

For each of these two methodologies, it is desirable to check that this methodology is *universal*, i.e., that if we use this methodology, then, for an arbitrary control function  $f : X \rightarrow Y$ , and for an arbitrary accuracy, there exist appropriate if-then rules for which the resulting control strategy represented by  $\tilde{f}(x)$  approximates the original control function  $f(x)$  within a given accuracy.

There exists many universal approximation results for approximations which are derived from Mamdani-style DNF formulas; first such results were formulated and proved, almost simultaneously, in 1990–92 papers by J. Buckley, Z. Cao, E. Czogala, D. Dubois, M. Grabisch, J. Han, Y. Hayashi, C.-C. Jou, A. Kandel, B. Kosko, J. Mendel, H. Prade, and L.-X. Wang; for a recent survey, see, e.g., [6] and references therein. There also exist several universal approximation results for implication-style CNF formulas [1, 2, 3, 11, 13].

These results are usually proved separately and provide two different (seemingly unrelated) approximations. In logic, however, CNF and DNF forms are related. In classical (2-valued) logic, every propositional formula  $F$  can be represented in both DNF and CNF forms  $F_{\text{DNF}}$  and  $F_{\text{CNF}}$ ; for every input  $x$ , these forms lead to exactly the same truth value:  $F_{\text{CNF}}(x) = F(x) = F_{\text{DNF}}(x)$ . In fuzzy logic, each propositional formula can also be transformed (generally, non-equivalently, see [11]) into CNF and DNF forms, so that using  $f_{\&} = \min$ ,  $f_{\vee} = \max$ , and  $f_{\neg}(a) = 1 - a$ , we get  $F_{\text{CNF}}(x) \leq F_{\text{DNF}}(x)$  (to be more precise,  $F_{\text{CNF}}(x) \leq F(x) \leq F_{\text{DNF}}(x)$ ; see, e.g., [12, 14]). In view of this relation, it is desirable to have a universal approximation result for CNF and DNF formulas which is consistent with this “fuzzy duality”, i.e., in which there is a similar relation between the fuzzy relations  $R_{\text{DNF}}(x, y)$  and  $R_{\text{CNF}}(x, y)$  which approximate the desired function  $f$ . Such a result is presented in this paper.

In proving this duality-related result, we also somewhat generalize the known CNF and DNF universal approximation theorems.

## 2 General Case: Functions Defined on an Arbitrary Compact Set

**Definition 1.** Let  $k = 1$  or  $k = 2$ , and let  $\oplus$  be a propositional  $k$ -ary operation in classical (2-valued) logic (e.g.,  $\&$ ,  $\vee$ ,  $\neg$ ,  $\rightarrow$ ). We say that an operation  $f_{\oplus} : [0, 1]^k \rightarrow [0, 1]$  is consistent with classical logic if it coincides with  $\oplus$  when all its inputs are 0's and 1's (corresponding to “false” and “true”).

Please note that we did not require that  $f_{\oplus}$  is continuous (as a function), or that  $f_{\&}(a, b)$  is commutative or associative, etc.

**Definition 2.** Let  $X$  be a compact metric space with a metric  $d_X$ ,  $Y$  be a complete metric space with a metric  $d_Y$ ,  $f : X \rightarrow Y$  be a continuous function from  $X$  to  $Y$ , and  $\varepsilon > 0$  be a real number. We say that a fuzzy relation  $R : X \times Y \rightarrow [0, 1]$   $\varepsilon$ -approximates a function  $f : X \rightarrow Y$  if the following two conditions hold:

- for every  $x \in X$ ,  $R(x, f(x)) > 0$ , and
- for every  $x \in X$  and  $y \in Y$ , if  $R(x, y) > 0$ , then  $d_Y(y, f(x)) \leq \varepsilon$ .

**Theorem 1.** Let operations  $f_{\&}$ ,  $f_{\vee}$ , and  $f_{\neg}$  be consistent with classical logic. Then, for every compact metric space  $X$ , for every continuous function  $f : X \rightarrow Y$  into a complete metric space  $Y$ , and for every real number  $\varepsilon > 0$ , there exist fuzzy rules of type (1) for which:

- both fuzzy relations  $R_{\text{DNF}}$  and  $R_{\text{CNF}}$  (obtained using the interpretation determined by  $f_{\&}$ ,  $f_{\vee}$ , and  $f_{\neg}$ )  $\varepsilon$ -approximate  $f$ , and
- $R_{\text{CNF}}(x, y) \leq R_{\text{DNF}}(x, y)$  for all  $x$  and  $y$ .

*Comment.* For the convenience of the readers, all the proofs are placed in the special Proofs section.

A similar result holds for  $R_{\rightarrow}$  instead of  $R_{\text{CNF}}$ :

**Theorem 1'.** Let operations  $f_{\&}$ ,  $f_{\vee}$ , and  $f_{\rightarrow}$  be consistent with classical logic. Then, for every compact metric space  $X$ , for every continuous function  $f : X \rightarrow Y$  into a complete metric space  $Y$ , and for every real number  $\varepsilon > 0$ , there exist fuzzy rules of type (1) for which:

- both fuzzy relations  $R_{\rightarrow}$  and  $R_{\text{DNF}}$  (obtained using the interpretation determined by  $f_{\&}$ ,  $f_{\vee}$ , and  $f_{\rightarrow}$ )  $\varepsilon$ -approximate  $f$ , and
- $R_{\rightarrow}(x, y) \leq R_{\text{DNF}}(x, y)$  for all  $x$  and  $y$ .

From the fact that a fuzzy relation  $R(x, y)$   $\varepsilon$ -approximates a function  $f(x)$ , we can conclude that the result  $\tilde{f}(x) = D(\mu_x)$  of applying, for every  $x \in X$ ,

a defuzzification procedure  $D$  (see below) to the corresponding membership function  $\mu_x(y) = R(x, y)$  is  $\varepsilon$ -close to  $f(x)$ :

**Definition 3.** By a defuzzification procedure, we mean a mapping  $D : [0, 1]^Y \rightarrow Y$  which maps every membership function  $\mu : Y \rightarrow [0, 1]$  (which is not identically zero) into an element  $D(\mu) \in Y$  for which  $\mu(D(\mu)) > 0$ .

**Proposition 1.** If a fuzzy relation  $R(x, y)$   $\varepsilon$ -approximates a function  $f(x)$ , then, for every defuzzification procedure  $D$ , the result  $\tilde{f}(x) = D(\mu_x)$  of applying this defuzzification procedure  $D$  to the corresponding membership function  $\mu_x(y) = R(x, y)$  is  $\varepsilon$ -close to  $f(x)$ , i.e.,  $d_Y(\tilde{f}(x), f(x)) \leq \varepsilon$ .

**Corollary 1.** Let operations  $f_{\&}$ ,  $f_{\vee}$ , and  $f_{\neg}$  be consistent with classical logic. Then, for every compact metric space  $X$ , for every continuous function  $f : X \rightarrow Y$  into a complete metric space  $Y$ , and for every real number  $\varepsilon > 0$ , there exist fuzzy rules of type (1) for which, for each defuzzification procedure  $D$ , the results  $\tilde{f}_{\text{DNF}}(x)$  and  $\tilde{f}_{\text{CNF}}(x)$  of defuzzifying the relations  $R_{\text{DNF}}$  and  $R_{\text{CNF}}$  (obtained using  $f_{\&}$ ,  $f_{\vee}$ , and  $f_{\neg}$ ) are  $\varepsilon$ -close to  $f$ .

**Corollary 1'.** Let operations  $f_{\&}$ ,  $f_{\vee}$ , and  $f_{\rightarrow}$  be consistent with classical logic. Then, for every compact metric space  $X$ , for every continuous function  $f : X \rightarrow Y$  into a complete metric space  $Y$ , for every real number  $\varepsilon > 0$ , there exist fuzzy rules of type (1) for which, for each defuzzification procedure  $D$ , the results  $\tilde{f}_{\text{DNF}}(x)$  and  $\tilde{f}_{\text{CNF}}(x)$  of defuzzifying the relations  $R_{\text{DNF}}$  and  $R_{\text{CNF}}$  (obtained using  $f_{\&}$ ,  $f_{\vee}$ , and  $f_{\rightarrow}$ ) are  $\varepsilon$ -close to  $f$ .

### 3 Towards a More Realistic Situation: Case When $Y = \mathbb{R}$

In the case when  $Y$  is a real line ( $Y = \mathbb{R}$ ), we can use a different class of possible “defuzzification procedures” and still get the same universal approximation result. Namely, we can use the following definition:

**Definition 3'.** ( $Y = \mathbb{R}$ ) By a defuzzification procedure, we mean a mapping  $D$  which maps every non-zero membership function  $\mu : \mathbb{R} \rightarrow [0, 1]$  into a real number  $D(\mu)$  in such a way that for an arbitrary interval  $[a, b]$ , if a membership function  $\mu(x)$  is equal to 0 for all values  $x$  outside an interval  $[a, b]$ , then  $D(\mu) \in [a, b]$ .

*Comment.* Both centroid and center-of-maximum are defuzzification procedures in this sense.

**Proposition 1'.** ( $Y = \mathbb{R}$ ) If a fuzzy relation  $R(x, y)$   $\varepsilon$ -approximates a function  $f(x)$ , then, for every defuzzification procedure  $D$ , the result  $\tilde{f}(x) = D(\mu_x)$  of applying this defuzzification procedure  $D$  to the corresponding membership function  $\mu_x(y) = R(x, y)$  is  $\varepsilon$ -close to  $f(x)$ , i.e.,  $d_Y(\tilde{f}(x), f(x)) \leq \varepsilon$ .

## 4 Realistic Case: Functions From $\mathbb{R}^n$ to $\mathbb{R}$

For the case when  $X = \mathbb{R}^n$ , we can prove similar results with rules of type (2):

**Theorem 2.** *Let operations  $f_{\&}$ ,  $f_{\vee}$ , and  $f_{\neg}$  be consistent with classical logic. Then, for every integer  $n > 0$ , for every compact set  $X \subset \mathbb{R}^n$ , for every continuous function  $f : X \rightarrow \mathbb{R}$ , and for every real number  $\varepsilon > 0$ , there exist fuzzy rules of type (2) for which:*

- both fuzzy relations  $R_{\text{DNF}}$  and  $R_{\text{CNF}}$  (obtained using  $f_{\&}$ ,  $f_{\vee}$ , and  $f_{\neg}$ )  $\varepsilon$ -approximate  $f$ , and
- $R_{\text{CNF}}(x, y) \leq R_{\text{DNF}}(x, y)$  for all  $x = (x_1, \dots, x_n) \in X$  and  $y$ .

**Theorem 2'.** *Let operations  $f_{\&}$ ,  $f_{\vee}$ , and  $f_{\rightarrow}$  be consistent with classical logic. Then, for every integer  $n > 0$ , for every compact set  $X \subset \mathbb{R}^n$ , for every continuous function  $f : X \rightarrow \mathbb{R}$ , and for every real number  $\varepsilon > 0$ , there exist fuzzy rules of type (2) for which:*

- both fuzzy relations  $R_{\rightarrow}$  and  $R_{\text{DNF}}$  (obtained using  $f_{\&}$ ,  $f_{\vee}$ , and  $f_{\rightarrow}$ )  $\varepsilon$ -approximate  $f$ , and
- $R_{\rightarrow}(x, y) \leq R_{\text{DNF}}(x, y)$  for all  $x(x_1, \dots, x_n) \in X$  and  $y$ .

In our universal approximation result, we prove that for every function  $f : X \rightarrow Y$ , we can select the rules and the membership functions for which we get the desired approximation property.

For Mamdani's (DNF) case, a stronger statement is true: that whatever "realistic" membership function  $\mu_0(x)$  we choose, we can always find rules in which all the membership functions  $A_{ik}(x_k)$  and  $B_i(y)$  are *of the type*  $\mu_0$ , i.e., they all have the form  $\mu(x) = \mu_0(a \cdot x + b)$  for some real numbers  $a \neq 0$  and  $b$  (see, e.g., [7, 8]).

From the proof of Theorems 2 and 2', we see that all the membership functions used in the approximation have the same type, i.e., that there is a type  $\mu_0$  which provides a universal approximation property both for DNF and for CNF forms. We do not know whether a similar result is true for an arbitrary given type.

## 5 Proofs

### 5.1 Proof of Theorem 1

This proof is similar to the original Kosko's proof [5] of a universal approximation result for DNF (i.e., for Mamdani methodology), and to our own proofs from [7, 8, 11].

1°. Let us take  $\varepsilon_1 = \varepsilon/2$ . Since a function  $f$  is continuous on a compact set  $X$ , it is also uniformly continuous. Therefore, there exists  $\delta > 0$  such that if  $d_X(x, x') \leq \delta$ , then  $d_Y(f(x), f(x')) \leq \varepsilon_1$ .

Since  $X$  is a compact metric space, there exists a finite  $\delta$ -net for  $X$ , i.e., a finite set of elements  $x^{(1)}, \dots, x^{(N)} \in X$  for which, for every  $x \in X$ , there exists an  $i$  for which  $d_X(x, x^{(i)}) \leq \delta$ . For each of these elements  $x^{(i)}$ , we can find  $y^{(i)} = f(x^{(i)})$ . We will show that Theorem 1 holds for  $N$  rules of type (1) where for every  $i$ ,

- $A_i(x) = 1$  if  $d_X(x, x^{(i)}) \leq \delta$ , and  $A_i(x) = 0$  otherwise;
- $B_i(y) = 1$  if  $d_Y(y, y^{(i)}) \leq \varepsilon_1$ , and  $B_i(y) = 0$  otherwise.

All these fuzzy sets  $A_i$  and  $B_i$  are crisp: indeed,  $A_i(x)$  is a characteristic function of the inequality  $d_X(x, x^{(i)}) \leq \delta$ , and  $B_i(y)$  is a characteristic function of the inequality  $d_Y(y, y^{(i)}) \leq \varepsilon_1$ .

Since all the  $f_{\&}$ ,  $f_{\vee}$ , and  $f_{\neg}$  are consistent with classical logic and  $A_i$ ,  $B_i$  are crisp, we conclude that the relations  $R_{\text{CNF}}$  and  $R_{\text{DNF}}$  can be obtained using classical logical connectives. Namely,

$$R_{\text{DNF}}(x, y) = 1 \iff \exists i \left( d_X(x, x^{(i)}) \leq \delta \ \& \ d_Y(y, y^{(i)}) \leq \varepsilon_1 \right); \quad (15)$$

$$R_{\text{CNF}}(x, y) = 1 \iff \forall i \left( d_X(x, x^{(i)}) > \delta \ \vee \ d_Y(y, y^{(i)}) \leq \varepsilon_1 \right). \quad (16)$$

2°. Let us first show that the relation  $R_{\text{DNF}}$   $\varepsilon$ -approximates the given function  $f$ .

2.1°. In accordance with the definition of  $\varepsilon$ -approximation, we first prove that for every  $x \in X$ , we have  $R_{\text{DNF}}(x, f(x)) > 0$ .

Indeed, let  $x$  be an arbitrary element of the set  $X$ . Since  $x^{(1)}, \dots, x^{(N)}$  is a  $\delta$ -net, there exists an  $i$  for which  $d_X(x, x^{(i)}) \leq \delta$ . Due to our choice of  $\delta$ , we conclude that  $d_Y(f(x), y^{(i)}) \leq \varepsilon_1$ . Thus, (15) is true, hence,  $R_{\text{DNF}}(x, f(x)) > 0$ .

2.2°. Let us now prove that for every  $x \in X$  and  $y \in Y$ , if  $R_{\text{DNF}}(x, y) > 0$ , then  $d_Y(y, f(x)) \leq \varepsilon$ .

Indeed, since  $R_{\text{DNF}}$  is a crisp relation, the only possibility for  $R_{\text{DNF}}(x, y) > 0$  is to have  $R_{\text{DNF}}(x, y) = 1$ , i.e.,  $R_{\text{DNF}}(x, y)$  to be true. This means that there exists an  $i$  for which  $d_X(x, x^{(i)}) \leq \delta$  and  $d_Y(y, y^{(i)}) \leq \varepsilon_1$ . Due to our choice of  $\delta$ , from  $d_X(x, x^{(i)}) \leq \delta$ , we can conclude that  $d_Y(f(x), f(x^{(i)})) = d_Y(f(x), y^{(i)}) \leq \varepsilon_1$ . Thus, from the triangle inequality, we conclude that  $d_Y(y, f(x)) \leq d_Y(y, y^{(i)}) + d_Y(y^{(i)}, f(x)) \leq \varepsilon_1 + \varepsilon_1 = \varepsilon$ . The statement is proven.



3°. Let us now show that the relation  $R_{\text{CNF}}$  also  $\varepsilon$ -approximates the given function  $f$ .

3.1°. In accordance with the definition of  $\varepsilon$ -approximation, we first prove that for every  $x \in X$ , we have  $R_{\text{CNF}}(x, f(x)) > 0$ .

Indeed, let  $x$  be an arbitrary element of the set  $X$ . For every  $i$ , we have  $d_X(x, x^{(i)}) \leq \delta$  or  $d_X(x, x^{(i)}) > \delta$ . By definition of  $\delta$ , the inequality  $d_X(x, x^{(i)}) \leq \delta$  implies that  $d_Y(f(x), f(x^{(i)})) = d_Y(f(x), y^{(i)}) \leq \varepsilon_1$ . Therefore, we have  $d_Y(f(x), y^{(i)}) \leq \varepsilon_1$  or  $d_X(x, x^{(i)}) > \delta$ . Hence, (16) is true, and  $R_{\text{CNF}}(x, f(x)) > 0$ .

3.2°. Let us show that for every  $x \in X$  and  $y \in Y$ , if  $R_{\text{CNF}}(x, y) > 0$ , then  $d_Y(y, f(x)) \leq \varepsilon$ .

Indeed, since  $R_{\text{CNF}}$  is a crisp relation, the only possibility for  $R_{\text{CNF}}(x, y) > 0$  is to have  $R_{\text{CNF}}(x, y) = 1$ , i.e.,  $R_{\text{CNF}}(x, y)$  to be true. This means that for every  $i$ , either  $d_X(x, x^{(i)}) > \delta$  or  $d_Y(y, y^{(i)}) \leq \varepsilon_1$ . Equivalently, this disjunction means that the inequality  $d_X(x, x^{(i)}) \leq \delta$  implies (in the logical sense, where implication is material!) that  $d_Y(y, y^{(i)}) \leq \varepsilon_1$ .

Since  $x^{(1)}, \dots, x^{(N)}$  is a  $\delta$ -net, there exists an  $i$  for  $d_X(x, x^{(i)}) \leq \delta$ . For this  $i$ , we thus have  $d_Y(y, y^{(i)}) \leq \varepsilon_1$ . On the other hand, due to our choice of  $\delta$ , from  $d_X(x, x^{(i)}) \leq \delta$ , we can conclude that  $d_Y(f(x), f(x^{(i)})) = d_Y(f(x), y^{(i)}) \leq \varepsilon_1$ . Thus, from the triangle inequality, we conclude that  $d_Y(y, f(x)) \leq d_Y(y, y^{(i)}) + d_Y(y^{(i)}, f(x)) \leq \varepsilon_1 + \varepsilon_1 = \varepsilon$ . The statement is proven.

4°. To complete the proof of the theorem, we must now show that  $R_{\text{CNF}}(x, y) \leq R_{\text{DNF}}(x, y)$  for all  $x$  and  $y$ .

Since both relations  $R_{\text{CNF}}(x, y)$  and  $R_{\text{DNF}}(x, y)$  are crisp, the desired inequality is equivalent to saying that for every  $x$  and  $y$ , if  $R_{\text{CNF}}(x, y)$  is true, then  $R_{\text{DNF}}(x, y)$  should also be true. Indeed, let  $R_{\text{CNF}}(x, y)$  hold for some  $x$  and  $y$ . This means that for every  $i$ , the inequality  $d_X(x, x^{(i)}) \leq \delta$  implies (in the logical sense) that  $d_Y(y, y^{(i)}) \leq \varepsilon_1$ . Since  $x^{(1)}, \dots, x^{(N)}$  is a  $\delta$ -net, there exists an  $i$  for which  $d_X(x, x^{(i)}) \leq \delta$ . Thus, for this  $i$ , we have  $d_Y(y, y^{(i)}) \leq \varepsilon_1$ . So, for this  $i$ , both inequalities  $d_X(x, x^{(i)}) \leq \delta$  and  $d_Y(y, y^{(i)}) \leq \varepsilon_1$  are true, hence the formula (16) is true. The statement is proven, and so is the theorem.

*Comment.* We have already mentioned that in crisp (2-valued) propositional logic, CNF and DNF forms represent the same function. It deserves mentioning that although in our proof, the relations  $R_{\text{CNF}}(x, y)$  and  $R_{\text{DNF}}(x, y)$  are both crisp, they do not necessarily the same. Indeed, let us consider the simplest case when  $X = Y = [0, 1]$  with a normal metric  $d_X(x, x') = d_Y(x, x') = |x - x'|$ , and  $f(x) = x$ . Then,  $\delta = \varepsilon_1 = \varepsilon/2$ . As a  $\delta$ -net, we can select the points  $x^{(i)} = (2i - 1) \cdot \delta$ , i.e.,  $x^{(1)} = \delta$ ,  $x^{(2)} = 3\delta$ , etc.; then,  $y^{(i)} = x^{(i)}$ . Here, for  $x = 2\delta$  and  $y = 0$ , we have  $d_X(x, x^{(1)}) \leq \delta$  and  $d_Y(y, y^{(1)}) \leq \varepsilon_1$ , so  $R_{\text{DNF}}(x, y)$

is true. However, the property  $R_{\text{CNF}}(x, y)$  is not true, because for  $i = 2$ , we have  $d_X(x, x^{(2)}) \leq \delta$ , but  $d_Y(y, y^{(2)}) = 3\delta = 3\varepsilon_1 > \varepsilon_1$ .

## 5.2 Proof of Theorem 1'

For this proof, we can use the exact same crisp sets  $A_i$  and  $B_i$  as in the proof of Theorem 1.

## 5.3 Proof of Proposition 1

By definition,  $\tilde{f}(x) = D(\mu_x)$ , where the membership function  $\mu_x : Y \rightarrow [0, 1]$  is defined as  $\mu_x(y) = R(x, y)$ . By the definition of a defuzzification procedure, for every  $x \in X$ , we have  $\mu_x(D(\mu_x)) > 0$ , i.e., by definition of  $\mu_x$ ,  $R(x, \tilde{f}(x)) > 0$ . From the definition of  $\varepsilon$ -approximation, we can now conclude that  $d_Y(\tilde{f}(x), f(x)) \leq \varepsilon$ . The proposition is proven.

## 5.4 Proof of Proposition 1'

By definition,  $\tilde{f}(x) = D(\mu_x)$ , where the membership function  $\mu_x : Y \rightarrow [0, 1]$  is defined as  $\mu_x(y) = R(x, y)$ . From the definition of  $\varepsilon$ -approximation, we conclude that if  $R(x, y) > 0$ , then  $d_Y(y, f(x)) = |y - f(x)| \leq \varepsilon$ . Thus, if  $|y - f(x)| > \varepsilon$ , we have  $R(x, y) = \mu_x(y) = 0$ . Hence, the function  $\mu_x(y)$  is equal to 0 outside the interval  $[f(x) - \varepsilon, f(x) + \varepsilon]$ . By definition of a defuzzification procedure, we can now conclude that the result  $\tilde{f}(x)$  of its defuzzification also belongs to the same interval, i.e., that  $|\tilde{f}(x) - f(x)| \leq \varepsilon$ . The proposition is proven.

## 5.5 Proof of Theorems 2 and 2'

This proof is similar to the proofs of Theorems 1 and 1'. Indeed, for  $X \subset \mathbb{R}^n$ , continuity of a function  $f : X \rightarrow \mathbb{R}$  with respect to a normal (Euclidean) metric is equivalent to its continuity with respect to the uniform metric  $d_X(x, x') = \max_i |x_i - x'_i|$ . Thus, from the proofs of Theorems 1 and 1', we conclude that there exist appropriate rules of type (1), with  $A_i(x) = 1 \iff d_X(x, x^{(i)}) \leq \delta$  and  $A_i(x) = 0$  for all other  $x$ , and  $B_i(y) = 1 \iff |y - y^{(i)}| \leq \varepsilon_1$  and  $B_i(y) = 0$  for all other  $y$ . By the definition of the uniform metric  $d_X$ , the inequality  $d_X(x, x^{(i)}) \leq \delta$  is equivalent to  $A_{i1}(x_1) \& \dots \& A_{in}(x_n) = 1$ , where  $A_{ik}(x_k) = 1 \iff |x_k - x_k^{(i)}| \leq \delta$  and  $A_{ik}(x_k) = 0$  for all other  $x_k$ . Thus, rules of type (1) can be reformulated in the desired form (2). The theorems are thus proven.

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