



Fuzzy Temporal Non-Monotonic Reasoning

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January 15, 2024

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Abstract— Non-monotonic reasoning is reasoning in which many conclusions will be drawn. The non-monotonic problem is undecidable. An undecided problem has no solution. Fuzzy logic will make undecided problems into decidable problems. Some time, fuzzy non-monotonic reasoning deals with fuzzy temporal constraints like pat, present feature etc. A young bird can fly but a very young bird may not fly. In this paper, fuzzy temporal non-monotonic reasoning is studied with unknown and known twofold fuzzy sets to make undecided problems into decidable. Fuzzy temporal truth maintenance system (FTTMS) is studied for computation of fuzzy non-monotonic reasoning. Some examples are given.

Keywords—non-monotonic reasoning, fuzzy sets, twofold fuzzy sets, fuzzy non-monotonic reasoning, FTTMS, incomplete knowledge

I. INTRODUCTION

Non-monotonic reasoning will draw different conclusions and it is undecided reasoning. Sometimes Artificial Intelligence (AI) has to deal with undecided problems. Non-monotonic problems are undecided. In non-monotonic reasoning, if some knowledge is added to the system then the conclusion will be changed, if the knowledge base is incomplete then the reasoning is also incomplete. Knowledge bases collect facts and give conclusions based on facts. If the knowledge is temporal, then the reasoning also changes. For instance, a young bird is kept in a room. After some time it is escaped. Whether the bird can fly or not? Here the age of the bird is involved.

John McCarthy [5] formalized non-monotonic reasoning with predicates $P(x_1, x_2, \dots, x_n)$ for the propositions of type “ x is A ”.

The non-monotonic logic may be defined as

$$\forall x (P(x) \wedge Q(x) \rightarrow R(x))$$

$$\exists P(x) \wedge Q(x) \rightarrow \neg R(x)$$

For instance,

$$\forall x (\text{bird}(x) \wedge \text{wings}(x) \rightarrow \text{fly}(x))$$

$$\text{bird}(\text{peacock}) \wedge \text{wings}(\text{peacock}) \rightarrow \text{fly}(\text{peacock})$$

$$\exists x (\text{bird}(x) \wedge \text{wings}(x) \rightarrow f) \rightarrow \neg \text{fly}(x)$$

$$\exists x (\text{bird}(\text{penguin}) \wedge \text{wings}(\text{penguin}) \rightarrow \neg \text{fly}(\text{penguin}))$$

I. FUZZY LOGIC

Zadeh [15] fuzzy logic is based on belief rather than probability (likelihood). The fuzzy logic makes imprecise information into precise.

Zadeh fuzzy logic is characterized by single membership function.

Zadeh [16] has introduced fuzzy sets as a model to deal with incomplete information. A fuzzy set is a class of objects with a continuum of grades.

The set A of X is characterized by its membership function $\mu_A(x)$ and ranging values in the unit interval $[0, 1]$

$\mu_A(x): X \rightarrow [0, 1], x \in X$, where X is the universe of discourse.

$A = \mu_A(x_1)/x_1 + \mu_A(x_2)/x_2 + \dots + \mu_A(x_n)/x_n$, where “+” is union

$$\mu_{\text{bird}}(x) = \mu_{\text{bird}}(x_1)/x_1 + \mu_{\text{bird}}(x_2)/x_2 + \dots + \mu_{\text{bird}}(x_n)/x_n$$

$$\mu_{\text{bird}}(x) = \mu_{\text{bird}}(x_1)/x_1 + \mu_{\text{bird}}(x_2)/x_2 + \dots + \mu_{\text{bird}}(x_n)/x_n$$

$$\mu_{\text{bird}}(x) = 0.1/\text{Penguin} + 0.3/\text{Hen} + 0.5/\text{Cock} + 0.6/\text{Parrot} + 0.8/\text{eagle} + 1.0/\text{flamingos}$$

$$\mu_{\neg A}(x) = 1 - \mu_A(x)$$

$$\mu_{A \wedge B}(x) = \min\{\mu_A(x), \mu_B(x)\}$$

$$\mu_{A \vee B}(x) = \max\{\mu_A(x), \mu_B(x)\}$$

$$\mu_{QA}(x) = \mu_A(x)^q$$

Negation

Conjunction

Disjunction

Quantifier

The fuzzy rules are of the form “if <Precedent Part> then <Consequent Part>”

if x is P then x is Q .

if x is P_1 and x is $P_2 \dots x$ is P_n then x is Q

The Zadeh [14] fuzzy conditional inference is given by

if x is P_1 and x is $P_2 \dots x$ is P_n then x is $Q =$

$$\min\{1, (1 - \min\{\mu_{P_1}(x), \mu_{P_2}(x), \dots, \mu_{P_n}(x)\}) + \mu_Q(x)\} \quad (2.1)$$

The Mamdani [8] fuzzy conditional inference is given by

if x is P_1 and x is $P_2 \dots x$ is P_n then x is $Q =$

$$\min\{\mu_{P_1}(x), \mu_{P_2}(x), \dots, \mu_{P_n}(x), \mu_Q(x)\} \quad (2.2)$$

The fuzzy conditional inference may be derived “Consequent Part” from “Precedent Part” [15].

Fuzzy conditional inference is given by fuzzy conditional inference

$$\text{If } x \text{ is } P \text{ then } x \text{ is } Q = \{\mu_P(x)\}$$

if x is P_1 and x is $P_2 \dots x$ is P_n then x is $Q =$

$$\{\min\{\mu_{P_1}(x), \mu_{P_2}(x), \dots, \mu_{P_n}(x)\}\} \quad (2.3)$$

$$P(x) \wedge Q(x) \rightarrow R(x) = \{P(x) \wedge Q(x)\}$$

For instance,

$$x \text{ is bird} \wedge x \text{ has wings} \rightarrow x \text{ can fly}$$

$$= x \text{ is bird} \wedge x \text{ has wings}$$

II. FUZZY TEMPORAL NON-MONOTONIC REASONING

The temporal logic is logic with time constraints and time variables “ $t_1 - t_0$ ” like “before”, “meet”, “after”, where starting time t_0 and ending time t_1 .

The time constraints are necessary to deal with data [1, 4]. Fuzzy temporal logic should deal with incomplete information of time constraints.

A temporal variable is “t₁-t₀”, where t₀ is starting time and t₁ ending time.

For instance “past”=t₁-t₀, t₁<t₀

“Present”= t₁ approximately t₀

“feature”=t₁-t₀, t₁>t₀

A fuzzy temporal set is set of temporal variables with interval” t₁-t₀”.

Fuzzy Temporal non-monotonic reasoning May be formalized with fuzzy predicate for the proposition of type “ x is A”, when A is fuzzy set may be defined as

$$\forall x (P(x) \wedge T(x) \rightarrow R(x))$$

$$\forall x (P(x) \wedge T(x) \rightarrow \neg R(x))$$

For instance ,

$$\forall x (\text{bird}(x) \wedge \text{young}(x) \rightarrow \text{fly}(x))$$

$$\forall x (\text{bird}(x) \wedge \text{young}(x) \rightarrow \neg \text{fly}(x))$$

x is bird \wedge x is young \wedge x is unknown to fly \rightarrow x can fly

Suppose ,

x is bird \wedge x is young \wedge x is unknown to fly \rightarrow x can ‘t fly

or

x is bird \wedge x is young \wedge x is unknown to fly \rightarrow x can’t fly

For example,

Ozzie is bird \wedge Ozzie is young \wedge Ozzie is unknown to fly \rightarrow Ozzie can fly

Ozzie is bird \wedge Ozzie is young \wedge Ozzie is known to fly \rightarrow Ozzie can’t fly

Ozzie is bird \wedge Ozzie is young \wedge Ozzie is unknown to fly \rightarrow Ozzie can fly

The fuzzy temporal non-monotonic logic may be defined as

$$\forall x (P(x) \wedge Q(x) \wedge T(x) \wedge R(x) \rightarrow S(x))$$

$$\forall x (P(x) \wedge Q(x) \wedge T(x) \wedge R(x) \rightarrow \neg S(x))$$

For instance ,

$\forall x (\text{bird}(x) \wedge \text{wings}(x) \wedge \text{young}(x) \wedge \text{unknown-of-fly}(x) \rightarrow \text{fly}(x))$

$\forall x (\text{bird}(x) \wedge \text{wings}(x) \wedge \text{young}(x) \wedge \text{known-of-fly}(x) \rightarrow \neg \text{fly}(x))$

III. GENERALIZED FUZZY TEMPORAL NON-MONOTONIC REASONING

Zadeh [13] Proposed fuzzy set with single membership function. The two fold fuzzy set [12] will give more evidence than single membership function.

The fuzzy non-monotonic set may defined with two fold membership function using unknown and known

Definition: Given some Universe of discourse X, the proposition “ x is P” is defined as its two fold fuzzy membership function as

$$\mu_P(x) = \{\mu_P^{\text{unknown}}(x), \mu_P^{\text{known}}(x)\}$$

$$P = \{\mu_P^{\text{unknown}}(x), \mu_P^{\text{known}}(x)\}$$

Where P is Generalized fuzzy set and $x \in X$,

$$\mu_P(x)^{(\text{unknown}, \text{known})} \wedge \mu_Q(x)^{(\text{unknown}, \text{known})} \rightarrow \mu_S(x)^{(\text{unknown}, \text{known})}$$

where P,Q and S are twofold fuzzy set known, known}.

$$\mu_{\text{bird}}(x) \wedge \mu_{\text{wings}}(x) \rightarrow \mu_{\text{fly}}(x)$$

$$\mu_{\text{bird}}(x)^{(\text{unknown}, \text{known})} \wedge \mu_{\text{wings}}(x)^{(\text{unknown}, \text{known})} \rightarrow \mu_{\text{fly}}(x)^{(\text{unknown}, \text{known})}$$

The conflict of the incomplete information may be defined by fuzzy certainty factor(FCF).

FCF=(unknown- known)

$$\mu^{\text{FCF}}_P(x) \rightarrow [0,1], x \in X$$

$$\mu_{\text{bird}}^{\text{FCF}}(x) \rightarrow [0,1]$$

Where known and unknown are the fuzzy membership functions.

The fuzzy non-monotonic reasoning will bring uncertain knowledge in to certain knowledge.

$$\mu_{\text{bird}}(x)^{(\text{unknown}, \text{known})} \rightarrow \mu_{\text{fly}}(x)$$

$$\mu_P(x)^{(\text{unknown}, \text{known})} \rightarrow \mu_S(x)$$

$$S = \begin{cases} \mu_P^{\text{FCF}}(x) = 1 & \mu_P^{\text{FCF}}(x) \geq \alpha, \\ 0 & \mu_P^{\text{FCF}}(x) < \alpha \end{cases}$$

$$\text{fly} = \begin{cases} \mu_{\text{fly}}^{\text{FCF}}(x) = 1 & \mu_{\text{bird}}^{\text{FCF}}(x) \geq 0.56 \\ 0 & \mu_{\text{bird}}^{\text{FCF}}(x) < 0.5 \end{cases}$$

Bird = {0.2/penguin +0.3/Ozzie+ 0.8/parrot+ 0.9/waterfowl + 1.0/eagle , 0.1/penguin +0.2/Ozzie+ . 0.1/parrot+ 0.1/waterfowl + 0.1/eagle }

Wings= {0.1/penguin +0.3/Ozzie+ 0.7/parrot+ 0.8/waterfowl + 0.9/eagle , 0.0/penguin +0.1/Ozzie+ . 0.2/parrot+ 0.1/waterfowl + 0.1/eagle }

Definition: The two fold fuzzy set for the proposition of the type “x is P” is given by

$$\mu_P(x) = \{\mu_P^{\text{unknown}}(x), \mu_P^{\text{known}}(x)\}$$

$$\mu_{\text{bird}}^{\text{FCF}}(x) = \{\mu_{\text{bird}}^{\text{unknown}}(x), -\mu_{\text{bird}}^{\text{known}}(x)\}$$

$$= \{0.2/penguin +0.3/Ozzie+ 0.8/parrot+ 0.9/waterfowl + 1.0/eagle , 0.1/penguin +0.1/Ozzie+ . 0.1/parrot+ 0.1/waterfowl + 0.1/eagle \}$$

$$= \{0.1/penguin +0.2/Ozzie+ . 0.7/parrot+ 0.8/waterfowl + 0.9/eagle \}$$

$$\mu_{\text{wings}}^{\text{FCF}}(x) = \{\mu_{\text{wings}}^{\text{unknown}}(x), -\mu_{\text{wings}}^{\text{known}}(x)\}$$

$$\text{wings} = \{0.1/penguin +0.3/Ozzie+ 0.7/parrot+ 0.8/waterfowl + 0.9/eagle , 0.0/penguin +0.1/Ozzie+ . 0.2/parrot+ 0.1/waterfowl + 0.1/eagle \}$$

$$= \{0.1/penguin +0.2/Ozzie+ . 0.5/parrot+ 0.6/waterfowl + 0.8/eagle \}$$

$$\mu_{\text{young}}^{\text{FCF}}(x) = \{\mu_{\text{wings}}^{\text{unknown}}(x), -\mu_{\text{wings}}^{\text{known}}(x)\}$$

young = {0.3/penguin + 0.4/Ozzie + 0.7/parrot + 0.8/waterfowl + 0.9/eagle, 0.0/penguin + 0.1/Ozzie + 0.2/parrot + 0.3/waterfowl + 0.3/eagle}
 = {0.3/penguin + 0.4/Ozzie + 0.3/parrot + 0.5/waterfowl + 0.6/eagle}

Using (2.3), fuzzy inference is given by
 if x is P₁ and x is P₂ ... x is P_n then x is Q

if x is P₁ and x is P₂ ... x is P_n then x is Q₁
 Q₁ = P₁ ∧ P₂ ∧ ... ∧ P_n

The fuzzy temporal non-monotonic logic may be defined as

∀x (P(x) ∧ Q(x) ∧ T(x) ∧ R(x) → S(x))
 ∀x (P(x) ∧ Q(x) ∧ T(x) ∧ R(x) → ¬S(x))

For instance,

∀x (bird(x) ∧ wings(x) ∧ young(x) ∧ unknown-of-fly(x) → fly(x))
 ∀x (bird(x) ∧ wings(x) ∧ young(x) ∧ known-of-fly(x) → ¬fly(x))

The two statements combined with two fold fuzzy logic.

x is bird ∧ x has wings ∧ x is young → x can fly
 x can fly = x is bird ∧ x is young ∧ x has wings

μ_{bird}(x) ∧ μ_{wings}(x) ∧ μ_{young}(x) → μ_{fly}(x)
 {μ_{bird}^{unknown}(x), μ_{bird}^{known}(x)} ∧ {μ_{bird}^{unknown}(x), μ_{bird}^{known}(x)} ∧ {μ_{young}^{unknown}(x), μ_{young}^{known}(x)} = μ_{fly}(x)
 μ_{bird}^{FCF}(x) ∧ μ_{wings}^{FCF}(x) ∧ μ_{young}^{FCF}(x) → μ_{fly}(x)
 if x is P₁ and P₂ ... x is P_n then x is P₁ and P₂ ... x is P_n
 = min(μ_{P1}(x), μ_{P2}(x), ..., μ_{Pn}(x))
 μ_{fly}(x) = μ_{bird}^{FCF}(x) ∧ μ_{wings}^{FCF}(x)

μ_{bird}^{FCF}(x) = {0.1/penguin + 0.1/Ozzie + 0.7/parrot + 0.8/waterfowl + 0.9/eagle}

μ_{wings}^{FCF}(x) = {μ_{wings}^{unknown}(x), μ_{wings}^{known}(x)}
 = {0.1/penguin + 0.2/Ozzie + 0.5/parrot + 0.6/waterfowl + 0.8/eagle}
 μ_{young}^{FCF}(x) = {μ_{young}^{unknown}(x), μ_{young}^{known}(x)}
 = {0.3/penguin + 0.4/Ozzie + 0.3/parrot + 0.5/waterfowl + 0.6/eagle}

Fuzzy conditional inference "consequent part" may be derived from "precedent part".

x is bird ∧ x has wings ∧ x is young → x can fly
 x can fly = min { x is bird, x has wings }
 = {0.1/penguin + 0.2/Ozzie + 0.3/parrot + 0.5/waterfowl + 0.6/eagle}

The inference of "x can fly" for α ≥ 0.5 is given by
 = 1/waterfowl + 1/eagle
 Here fuzzy logic made incomplete information to precise information's. Some birds can fly and some birds can't fly.
 The fuzzy decision sets or quasi fuzzy set is defined by

$$R = \mu_A^R(x) = \begin{cases} 1 & \mu_A^{FCF}(x) \geq \alpha \\ 0 & \mu_A^{FCF}(x) < \alpha \end{cases}$$

The parrot, waterfowl and eagle can fly.

The penguin and Ozzie can't fly

The inference of "x can't fly" for α < 0.5 is given by
 = 0.1/penguin + 0.2/Ozzie

The inference of "x can fly" for α ≥ 0.5 is given by
 = 0.6/parrot + 0.6/waterfowl + 0.8/eagle

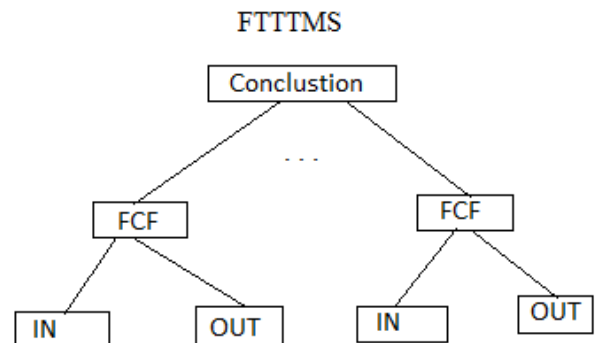
he parrot, waterfowl and eagle can fly and, penguin and Ozzie are can't fly.

IV. FUZZY TRUTH MAINTANACE SYSTEM

Doyel [4] studied truth maintenance system TMS] for non-monotonic reasoning

The fuzzy truth maintenance system (FTTMS) for fuzzy non-monotonic reasoning using fuzzy conditional inference as

FTTMS is having There is list of justification and conditions.



if x is bird and x has wings and x is young then x can fly

List L(IN-node, OUT-node), FCF-node

IN-node is unknown

OUT-node is known

FCF is (belirf-disbilief)

Consider the proposition "if x is bird then x can fly)

x is bird

IN 0.9

OUT 0.1

FCF 0.8

Conclusion : x can fly if FCF ≥ 0.5.

x is bird

IN 0.3

OUT 0.1
 FCF 0.2
 Conclusion :x can't fly if $FCF < 0.5$.

Consider the proposition “ if x has wings then x is fly)
 x has wings

IN 0.8
 OUT 0.1
 FCF 0.7
 Conclusion : x can fly if $FCF \geq 0.5$.

x has wings
 IN 0.3

OUT 0.1
 FCF 0.2
 Conclusion :x can't fly if $FCF < 0.5$.

Consider the proposition “ if x is young then x is fly)

X is young
 IN 0.6

OUT 0.1
 FCF 0.5
 Conclusion : x can fly if $FCF \geq 0.5$.

x is young
 IN 0.5

OUT 0.1
 FCF 0.4
 Conclusion :x can't fly if $FCF < 0.5$.

Consider the proposition “ if x is bird and x has wings and
 x is young then x is fly)

x is bird
 IN 0.9
 OUT 0.1
 FCF1 0.8
 x has wings
 IN 0.8
 OUT 0.1
 FCF2 0.7

X is young
 IN 0.6
 OUT 0.1
 FCF3 0.5

$$FCF = \min\{FCF1, FCF2, FCF3\} = 0.5$$

Conclusion : x can fly if $FCF \geq 0.5$.

X is very young
 FCF3 0.5

$$FCF = \min\{FCF1, FCF2, FCF3\} = 0.25$$

Conclusion :x can't fly (if $FC < 0.5$).

Here is the fuzzy temporal no monotonic logic is making imprecise to precise.

V. CONCLUSION

Non-monotonic reasoning will give different conclusions if some knowledge is added to the theory. Fuzzy non-

monotonic logic made imprecise to precise. A sometimes time constraint has to deal by fuzzy non-monotonic reasoning. A bird can't fly in the past but it can fly present and it can 'fly feature We have shown the difference between fuzzy temporal fuzzy non-monotonic reasoning and fuzzy monotonic reasoning..

ACKNOWLEDGMENT

The author thanks to Sri Venkateswara University for providing facilities to carry out this work.

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