

New Method of Fuzzy Conditional Inference and Approximate Reasoning

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Abstract

We consider fuzzy conditional inference of the form "if x is P then y is Q", "if x is P then y is Q else y is R" and "x is P_1 and/or P_2 and/or \cdots and/or P_n then y is R". In this paper, we propose method of fuzzy conditional inference.we applied the method on logical constructs developed by Fukami, Muzimoto and Tanaka. We have shown how our method satisfy intuitions under several criteria.

Keywords: Fuzzy logic, Fuzzy intuitions, Fuzzy conditional inference, Approximate reasoning

1. Introduction

Fukami, Mizumoto and Tanaka [2] developed logical constructs and shown that Zadeh [8] fuzzy conditional inference is not fit for intuitions. Fukami, Mizumoto and Tanaka adapted the Godel definition and Standard sequence methods to satisfy some fuzzy intuitions. The proposed method satisfy all the fuzzy intuitions proposed with fuzzy plausibility. We considered three criteria.

Criteria-1

If x is P then y is Qx is P_1

y is ?

If Apple is red then Apple is ripe apple is very ripe

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y is ?

Criteria-2 If x is P then y is Q else y is R

y is ?

x is P_1

If Apple is Ripe then Apple is Taste else Apple is Sour apple is very ripe

y is ?

Criteria-3

If x is P and x is Q or x is R then y is S x is P_1 and x is Q_1 or x is R_1

y is ?

If x is Red or x is ripe and x is big then x is taste x is red or x is ripe and x is very big

y is ?

1.1. Fuzzy Conditional Inference

A fuzzy set P is define by its characteristic function $\int \mu_P(x)/x$, $x \in X$, where x is individual and X is universe of discourse. $P = \int \mu_P(x)/x$ $P' = 1 - \int \mu_P(x)/x$ $P \lor Q = \max \{ (\int \mu_P(x), \int \mu_Q(y))(x, y) \}$ $P \land Q = \min \{ (\int \mu_P(x), \int \mu_Q(y))(x, y) \}$ $P \oplus Q = \min \{ 1, (\int \mu_P(x) + \int \mu_Q(y))(x, y) \}$ The fuzzy conditional propositions of the form "if (precedent part) then (consequent part)".

Consider the proposition of type "if x is P then y is Q "

Zadeh [18] fuzzy conditional inference is given as

 $\begin{array}{l} P \rightarrow Q = P' \oplus Q = \min \ \{ \ 1, \ 1 - (\int \mu_P(x) + \int \mu_S(y)) \} \\ \text{Consider the proposition of type "if x is P then y is Q else x is R".} \\ \text{It may be defined as "if x is P then y is Q \lor if x is P' then x is R"} \\ \text{It is given by} \\ \text{"if x is P then y is Q"} \\ \text{"if x is P' then x is R"} \end{array}$

Zadeh [8] fuzzy conditional inference is given as $P \to Q = P' \oplus Q = \min \{ 1, 1 - (\int \mu_P(x) + \int \mu_Q(y)) \}$ $P' \to R = P \oplus R = \min \{ 1, (\int \mu_P(x) + \int \mu_R(y)) \}$

1.2. Improved method

The consequent part is derived from precedent part for fuzzy conditional inference [5].

if x is A then y is B = A $\int \mu_B(y) = \int \mu_A(x)$, i.e., $B \subseteq A$ and $A \subseteq B$ (2.1) Consider fuzzy quantifiers A^{α} and B^{α} $A^{\alpha} \subseteq B$, i.e., $A^{\alpha} \leq B$ $B^{\alpha} \subseteq A$, i.e., $B^{\alpha} \leq A$

The fuzzy conditional inference is given by using Mamdani fuzzy conditional inference

if x is A then y is $B = \{A \times B\}$

The fuzzy conditional inference is give by using (2.1)

if x is A then y) is $B = \int \mu_B(y) \times \int \mu_A(x)$ if x is A then y is $B = \int \mu_B(y) \wedge \int \mu_A(x)$

if x is A then y is
$$B = \{ \int \mu_A(x), \int \mu_A(x) = \int \mu_B(y) \}$$
 (2.2)

2. Fuzzy Conditional Inference

2.1. Verification of Criteria-1

The fuzzy conditional inference may be given for Criteria-1 by

| Intuition | Proposition | Inference |
|-----------|-----------------------|-----------------------|
| I-1 | $x 	ext{ is } P$ | y is Q |
| I-2 | y is Q | x is P |
| II-1 | x is very P | y is very Q |
| II-2 | y is very Q | x is very P |
| III-1 | x is more or less P | y is more or less Q |
| III-2 | y is More or less Q | is more or less P |
| IV-1 | x is not P | y is not Q |
| IV-2 | y is not Q | x is not P |

Table 1: Criteria-1

Consider the fuzzy conditional inference

 $\int \mu_P(x)/x \to \int \mu_Q(y)/y = \{ \int \mu_P(x)/x \}$ and $\int \mu_P(x) = \int \mu_Q(y)$

2.2. Verification of fuzzy Intuitions for Criteria-1

2.1.1 In the case of Intuition I-1, II-1 and III-1

$$\begin{split} P^{\alpha} & o \ (\mathcal{P} \rightarrow \mathcal{Q}) \\ &= \int \mu_{\mathcal{P}^{\alpha}}(x) / x \ o \ (\int \mu_{\mathcal{P}}(x) / x \rightarrow \int \mu_{Q}(y) / y) \\ &= \int \mu_{\mathcal{P}^{\alpha}}(x) / x \ o \ (\int \mu_{\mathcal{P}}(x) / x) \\ &= \int \mu_{Q^{\alpha}}(y) / y \land (\int \mu_{Q}(y) / y \end{split}$$

$$\begin{split} &\int \mu_{Q^{\alpha}}(y)/y \\ &= y \text{ is } Q^{\alpha}(y)/y \\ &\text{Intuition I-1, II-1 and III-1 are satisfied.} \end{split}$$

2.1.2 In the case of Intuition I-2, II-2 and III-2

$$\begin{array}{l} (P \to \mathbf{Q}) \mathrel{\circ} Q^{\alpha} \\ = (\int \mu_P(x)/x \to \int \mu_Q(y)/y) \mathrel{\circ} \int \mu_{Q^{\alpha}}(y)/y \end{array}$$

 $= \int \mu_P(x)) \circ \int \mu_{Q^{\alpha}}(y)$ = $\int \mu_P(x)) \wedge \int \mu_{P^{\alpha}}(x)$

$$= \int \mu_{P^{\alpha}}(x)/x$$

x is P^{α}
Intuition I-2, II-2 and III-2 are satisfied.

2.1.3 In the case of Intuition IV-1 $\,$

 $\begin{array}{l} \mathrm{P' \ o \ }(\mathrm{P} \rightarrow \mathrm{Q}) \\ = \int \mu_{P'}(x)/x \ \mathrm{o \ }(\int \mu_{P}(x)/x \rightarrow \int \mu_{Q}(y)/y) \\ = \int \mu_{P'}(x)/x \ \mathrm{o \ }(\int \mu_{P}(x)/x) \\ = \int \mu_{Q}(y)/y \land (\int \mu_{Q}(y)/y \end{array}$

 $= \int \mu_{Q'}(y)/y$ y is not Q Intuition IV-1 satisfied.

2.1.4 In the case of Intuition IV-2

 $\begin{array}{l} (P \to \mathbf{Q}) \mathrel{\circ} \mathbf{Q}' \\ = (\int \mu_P(x)/x \to \int \mu_Q(y)/y) \mathrel{\circ} \int \mu_{Q'}(y)/y \\ = \int \mu_P(x)) \mathrel{\circ} \int \mu_Q'(x) \\ = \int \mu_P(x)) \mathrel{\wedge} \int \mu_{P_*}(x) \end{array}$

$$= \int \mu_{P.}(x)/x$$

x is not P
Intuition IV-2 satisfied.
Criteria-1 is satisfies I-1,I-2, II-1, II-2, III-1 and III-2, IV-1, IV-2.

2.3. Verification of Criteria-2

Consider the fuzzy conditional inference

$$\int \mu_{P'}(x)/x \to \int \mu_{R}(y)/y = \{\int \mu_{P'}(x)/x\}$$

and
$$\int \mu_{P'}(x) = \int \mu_{R}(y)$$

| The fuzzy conditional inference may | be given for Criteria-2 by |
|-------------------------------------|----------------------------|
|-------------------------------------|----------------------------|

| Intuition | Proposition | Inference |
|-----------|------------------------|-----------------------|
| I-1 | $x 	ext{ is } P$ | y is Q |
| I-2 | y is Q | x is P |
| II-1 | x is very P | y is very Q |
| II-2 | y is very Q | x is very P |
| III-1 | x is more or less P | y is more or less Q |
| III-2 | y is More or less Q | is more or less P |
| IV-2 | y is not R | x is not P |
| I'-1 | x is P' | y is R |
| I'-2 | y is R | x is P' |
| II'-1 | x is very P' | y is very R |
| II'-2 | y is very R | x is very P' |
| III'-1 | x is more or less P' | y is more or less R |
| III'-2 | y is More or less R | is more or less P' |
| IV'-2 | x is R' | y is P |

Table 2: Criteria-2

Criteria-1 is verified for I-1,I-2, II-1, II-2, III-1 and III-2, IV-2 in Criteria-1.

2.2.1 In the case of Intuition I'-1, II'-1 and III'-1

 $P^{\prime \alpha} \circ (\mathbf{P}^{\prime} \to \mathbf{R}) = \int \mu_{P^{\prime \alpha}}(x)/x \circ (\int \mu_{P^{\prime}}(x)/x \to \int \mu_{R}(y)/y) = \int \mu_{P^{\prime}}(x)/x \circ (\int \mu_{P^{\prime}}(x)/x) = \int \mu_{R^{\prime \alpha}}(y)/y \wedge \int \mu_{R}(y)/y = \int \mu_{R^{\prime \alpha}}(y)/y$

y is R'^α Intuition I'-1, II'-1 and III'-1 are satisfied.

2.2.2 In the case of Intuition I'-2, II'-2 and III'-2

 $\begin{array}{l} (\mathrm{P}' \to \mathrm{R}) \circ R^{\alpha} \\ = (\int \mu_{P'}(x)/x \to \int \mu_{R}(y)/y) \circ \int \mu_{R^{\alpha}}(y)/y \\ = \int \mu_{P'}(x)) \circ \int \mu_{R^{\alpha}}(y) \\ = \int \mu_{P'}(x))/x \wedge \int \mu_{R^{\alpha}}(x)/x \end{array}$

 $= \int \mu_{P'^{\alpha}}(x)/x$ x is R'^{α} Intuition I'-2, II'-2 and III'-2 are satisfied. 2.2.7 In the case of Intuition IV'-2

 $\begin{array}{l} (\mathbf{P}' \to \mathbf{R}) \mathbf{o} \ R'^{\alpha} \\ = (\int \mu_{P'}(x)/x \to \int \mu_R(y)/y) \ \mathbf{o} \ \int \mu_{R'^{\alpha}}(y)/y \\ = \int \mu_{P'}(x)) \ \mathbf{o} \ \int \mu_{R'^{\alpha}}(y) \\ = \int \mu_{P'}(x))/x \wedge \int \mu_{P^{\alpha}}(x)/x \end{array}$

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$$= \int \mu_{P^{\alpha}}(x)/x$$

x is P^{α}
Intuition IV'-2 satisfied.
Criteria-2 is satisfies I'-1,I'-2, II'-1, II'-2, IV'-2.

3. Verification of fuzzy Intuitions for Criteria-3

Consider fuzzy conditional inference

If x is P and x is Q or x is R then y is S x is P_1 and x is Q_1 or x is R_1

y is ?

Fuzzy inference is given by using Specialization and Generalization

If x is P then y is S x is P_1 y is ? If x is x is Q then y is S x is x is Q_1 y is ? If x is R then y is S x is R_1 y is ?

Fuzzy inference may be verified in the similar lines of Criteria-1

4. Business Application

The Business intelligence needs commonsense. The Business data is defied with fuzziness with linguistic variables.

For example

If x is Demand then Apple is Production apple is very Demand y is ?

If Apple is Sales then Price is Taste else Apple is Stock apple is very Sales

y is ?

If x is *Demand* or x is *Sales* and x is *Price* then y is *Production* x is more *Demand* or x is very *Sales* and x is *Price*

y is ?

These Criteria shall be studied with Criteria-1, Criteria-2 and Criteria-3.

5. Conclusion

In this paper, we consider the fizzy condition inference

If x is P then y is Q $x ext{ is } P_1$ $y ext{ is } P_1$ $y ext{ is } P$ then y is Q else y is R $x ext{ is } P_1$ $y ext{ is } P$ If x is P and x is Q or x is R then y is S $x ext{ is } P_1$ and x is Q_1 or x is R_1 $y ext{ is } P$

We try to prove three criteria with our method using fuzzy plausibility and it is approximate reasoning.

References

- E.H.Mamdani, Application of Fuzzy Logic to Approximate Reasoning Using Linguistic Synthesis. IEEE Trans. Computers. vol.26, no.12, pp.1182-1191, 1977.
- [2] S. Fukami, M. Muzumoto, K. Tanaka, Some Considerationas on Fuzzy Conditional Inference, Fuzzy Seta and Systems, vol.4, pp.243-273, 1980.
- [3] C. Howson, Successful Business Intelligence, McGraw-Hill, 2014.
- [4] N. Rescher, Many-valued Logic, McGrow-Hill, New York, 1969.
- [5] Poli Venkata Subba Reddy, M. Syam Babu, Some methods of reasoning for conditional propositions, Fuzzy Sets and Systems, vol.52, no.3, pp.229-250, 1992.
- [6] Poli Venkata Subba Reddy, Fuzzy conditional inference for medical diagnosis, Proceedings, Second International Conference on Fuzzy Theory and Technology, Durham, FT&T'93, vol.3, pp.193-195, 1993.
- [7] Poli Venkata subba reddy, Fuzzy logic based on Belief and Disbelief membership functions, Fuzzy Information and Engineering, Vol.9, no.4, pp.405-422, 2017.
- [8] L. A Zadeh, "Calculus of fuzzy Restrictions", In Fuzzy set s and their Applications to Cognitive and Decision Processes, L. A. Zadeh, King-Sun FU, Kokichi Tanaka and Masamich Shimura (Eds.), Academic Press, New York, pp.1-40, 1975.
- [9] L.A. Zadeh, Fuzzy sets. Information and Control, vol.8, pp.338-353., 1965.
- [10] A. Bochman, A logic for causal reasoning, Proceedings IJ-CAI2003, Morgan Kaufmann, 2003.