

Fuzzy Conditional Inference on Fuzzy Constructs: An Application to Data Analytics

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Abstract

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We consider fuzzy inference of the form "if \cdots then \cdots else \cdots " and "and/or ". We developed logical constructs based on logical intuitions developed by Fukami. With the propose method of fuzzy inference and causal logic , we apply on logical constructs. We try to show the fuzzy inference satisfy all intuitions under several criteria.

Keywords: fuzzy logic, fuzzy conditional inference, fuzzy intuitions, Complicated

sectionIntroduction Zadeh [7] and Mamdani [1] proposed the fuzzy conditional inference methods. Mizumoto [2] developed logical constructs and proved that Zadeh and Mamdani fuzzy conditional inference is not fit for intuitions. Mizumoto adapted the Godel definition and Standard sequence methods to prove some fuzzy intuitions. In the following, we developed logical constructs based on logical constructs developed by Muzumoto. We shown the our fuzzy inference method satisfy all the intuitions under several criteria. The fuzzy intuitions are studied for "and/or" and "if ··· then ··· else ···".

Type-1

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 Supply or x is Demand and x is Price then x is increase Profit x is more Supply or x is very Demand and x is more Price

y is ?

Type-2 Consider fuzzy inference

If x is P then y is Q else y is R x is P_1

y is ?

If x is Demand then x is Profit else x is Loss x is very Demand

y is ?

1. Some Methods of Fuzzy Conditional Inference

The fuzzy conditional propositions of the form "if (precedent part) then (consequent part)".

Mamdani [5] fuzzy conditional inference given as $A \to B == \{A \times B\}$. if x_1 is A_1 and x_2 is $A_2 \cdots x_n$ is A_n , then y is B $=\min\{\min(A_1, A_2, \cdots, A_n), B\}$

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

if x is A then y is B = A B = A is $B \subseteq A$ and $A \subseteq B$ if x_1 is A_1 and x_2 is $A_2 \cdots x_n$ is A_n then y is $B = x_1$ is A_1 and x_2 is $A_2 \cdots x_n$ is A_n

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

if x is A then y is $B = \{A \times B\}$ if x_1 is A_1 and x_2 is A_2 and \cdots and x_n is A_n then y is B={ $(A_1 \text{ and } A_2 \text{ and } \cdots \text{ and } A_n) \times (A_1 \text{ and } A_2 \text{ and } \cdots \text{ and } A_n)$ } = min{ A_1, A_2, \dots, A_n, B } The fuzzy conditional inference is give as [5]

if x_1 is A_1 and x_2 is A_2 and \cdots and x_n is A_n then y is $B = \{\min(A_1, A_2, \cdots, A_n)\}$ if x is A then y is $B = \{A\}$ (2.2)

2. Fuzzy Inference on Fuzzy Intuitions

Consider the causal logical inference [11] **Modus Pones** $P \rightarrow q$ P q **Modus Tollens** $p \rightarrow q$ q' q' p' **Generalization** $p \lor q = p$ $p \lor q = q$ **Specialization** $p \land q = p$ $p \land q = q$

Fuzzy plausibility

Plausibility theory will perform inconsistent information into consistent. Generalization

 $\begin{array}{l} p \lor \mathbf{q}, \, \mu = \mathbf{p}, \, \mu \\ p \lor \mathbf{q}, \, \mu = \mathbf{q}, \, \mu \end{array}$

Specialization

 $p \wedge \mathbf{q}$, $\mu {=} \mathbf{p}, \, \mu$

 $p \wedge \mathbf{q} = \mathbf{q}, \mu$

The inference is given using generalization and specialization $p \wedge q \lor r, \mu = p, \mu$

 $\begin{array}{l} p \wedge q \lor r, \mu = q, \mu \\ p \wedge q \lor r, \mu = r, \mu \end{array}$

Consider fuzzy inference Type-1

The fuzzy inference is given for Type-1 using generalization and specialization

If x is P then y is S x is P_1 y is S_1 If x is Q then y is S x is Q_1 y is S_1 If x is R then y is S x is R_1 y is S_1

Confider fuzzy inference Type-2

The fuzzy inference is given for Type-2 using generalization and specialization

If x is P then y is Q x is P_1 y is Q_1 If x is P' then y is R x is P_1 y is R_1 From further condition

From fuzzy conditional inference Type-1 and Type-2, the two criterions may be given as

Criteria-1

if x is P then y is S

x is P_1

y is textit S_1 **Criteria-2** if x is P' then y is Rx is P'_1

y is R_1

3.

Fuzzy Intuitions using New Fuzzy Inference

The fuzzy intuitions are give for Criteria-1.

Intuition	Proposition	Inference		
I-1	x is P	y is S		
I-2	y is S	x is P		
II-1	x is very P	y is very S		
II-2	y is very S	x is very P		
III-1	x is more or less P	y is more or less S		
III-2	y is More or less S	is more or less P		
IV-1	x is not P	y is not S		
IV-2	y is not S	x is not P		

Table 1: Fuzzy inference for Criteria-1

The fuzzy intuitions are give for Criteria-2.

x is P'	y is R
y is R'	x is P'
x is very P'	y is very R
y is very R	x is very P'
x is more or less P'	y is more or less R
y is More or less R	is more or less P'
x is not P'	y is not R
y is not R	x is not P'
	x is very P' $y is very R$ $x is more or less P'$ $y is More or less R$ $x is not P'$

Table 2: Fuzzy inference for Criteria-2

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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 \overline{y} is S_1 The inference is given using generalization and specialization If x is P then y is S x is P_1 \overline{y} is S_1 If x is Q then y is S x is Q_1 \overline{y} is S_1 If x is R then y is S x is R_1 \overline{y} is S_1

Fuzzy conditional inference may be given by combining inferences using Criteria-1

Consider fuzzy conditional inference Type-2

If x is P then y is Q else y is R $x ext{ is } P_1$ $y ext{ is } R_1$ If x is P then y is Q else y is R may be given by [4] $(if xisPthenyisQ) \lor (if xisP'thenxisR)$ $x ext{ is } P_1$ $y ext{ is } R_1$

If x is P then y is Q

x is P_1

y is R_1

Fuzzy conditional inference may be given as Criteria-1

If x is P' then y is R x is P_1 $\overline{\qquad}$ y is R_1

Fuzzy conditional inference may be given using criteria-1 or Criteria-2

4. Application to Fuzzy Intuitions

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

If x is Demand then y is Profit x is P_1

y is R_1

Consider the fuzzy data sets for production The fuzzy conditional infer-

Item No.	Demand
Item1	0.3
Item2	0.5
Item3	0.7
Item4	0.8
Item5	1.0

Table 3: Fuzzy data set Demand

ence using is given by

if x is Demand then x is Profit The fuzzy conditional inference using (3.1) given by T inference for Criteria-1 is given by

The fuzzy conditional

Item No.	Profit
Item1	0.3
Item2	0.5
Item3	0.7
Item4	0.8
Item5	1.0

Table 4: Fuzzy data set Profit

Table 5: Fuzzy conditional inference

Item No.	I-1	I-2	II-1	II-2	III-1	III-2	IV-1	IV-2
Item1	0.3	0.3	0.09	0.09	0.55	0.55	0.7	0.7
Item2	0.5	0.5	0.25	0.25	0.71	0.71	0.5	0.5
Item3	0.7	0.7	0.49	0.49	0.84	0.84	0.3	0.3
Item4	0.8	0.8	0.64	0.64	0.89	0.89	0.2	0.2
Item5	1.0	1.0	1.0	1.0	1.0	1.0	0.0	0.0

If x is not Demand then y is Loss x is P_1 _____

y is R_1

The fuzzy conditional inference using is given by

if x is not Demand then x is Loss

The fuzzy conditional inference using (3.1) given by

The fuzzy conditional

Item No.	Loss
Item1	0.7
Item2	0.5
Item3	0.3
Item4	0.2
Item5	0.0

Table 6: Fuzzy data set Loss

inference for Criteria-2 is given by

Table 7:	Fuzzy	$\operatorname{conditional}$	inference
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Item No.	I'-1	I'-2	II'-1	II'-2	III'-1	III'-2	IV'-1	IV'-2
Item1	0.7	0.7	0.49	0.49	0.84	0.84	0.7	0.7
Item2	0.5	0.5	0.25	0.25	0.71	0.71	0.5	0.5
Item3	0.3	0.3	0.09	0.09	0.84	0.84	0.3	0.3
Item4	0.2	0.2	0.04	0.04	0.55	0.55	0.3	0.3
Item5	0.0	0.0	0.0	0.0	0.0	0.0	1.0	1.0

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