

Approximate Fuzzy Reasoning

Kudaja

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- Introduction
- Modus Ponens in Fuzzy Logic
- Multicriteria Modus Ponens
- FATI and FITA approach to Multicriteria Modus Ponens
- CRI – Compositional rule of inference
- SBR – Similarity Based Reasoning
- Properties of Modus Ponens, visualization
- Other inference schemes for approximate reasoning

- What is „fuzzy logic“ ?
- What is „modus ponens“ ?
- How does one go from standard logic to fuzzy logic ?
- How do logic operands work and what is a t_norm,
co_norm ?

In standard logic, Modus Ponens can be expressed as:

$$B' = A' \wedge (A \implies B)$$

In fuzzy logic, Modus Ponens can be expressed as:

$$B'(y) = \sup_{x \in X} \left(\mathcal{T} \left(A'(x), \mathcal{I}(A(x), B(y)) \right) \right)$$

Where:

- A is the first statement (antecedent)
- A' is the fact
- B is the second statement (consequent)
- B' is the conclusion

Now lets define a criteria (rule) \mathcal{R} as:

$$\mathcal{R}(x, y) = \mathcal{I}(A(x), B(y))$$

In other words \mathcal{R} can be expressed as:

$$\mathcal{R} : \text{IF } x \text{ is } A \text{ THEN } y \text{ is } B$$

Now lets expand our criteria to a set of criterias as: \mathcal{R}_i as:

$$\mathcal{R}_i(x, y) = \mathcal{I}(A_i(x), B_i(y)), i = 1, 2, \dots, m$$

Or expressed as:

$$\mathcal{R}_i = \text{IF } x_1 \text{ is } A_{i1} \text{ AND } x_2 \text{ is } A_{i2} \text{ AND } \dots x_n \text{ is } A_{in} \text{ THEN } y \text{ is } B_i$$

$$A_{ij} \in \mathcal{F}(x_j) \text{ in rule } \mathcal{R}_i,$$

$$i = 1, 2, \dots, m,$$

$$j = 1, 2, \dots, n$$

FITA and FATI approach to Multicriteria Modus Ponens

- **FITA** - First Infer Then Aggregate

$$Z_{out} : B'(y) = \mathcal{T}\left(B'_1(y), B'_2(y), \dots, B'_m(y)\right)$$

$$B'(y) = \mathcal{T}_{i=1}^m(B'_i(y)), \text{ know: } B'_i(y) = \mathcal{T}\left(A'(x), \mathcal{R}_i(x, y)\right)$$

$$B'(y) = \mathcal{T}_{i=1}^m\left(\sup_{x \in X} \widehat{\mathcal{T}}(A'(x), \mathcal{R}_i(x, y))\right)$$

$$B'(y) = \mathcal{T}_{i=1}^m\left(\sup_{x \in X} \widehat{\mathcal{T}}\left(A'(x), \mathcal{I}(A_i(x), B_i(y))\right)\right)$$

FITA and FATI approach to Multicriteria Modus Ponens

- **FATI** - First Aggregate Then Infer

$$\widehat{\mathcal{R}}(x, y) = \mathcal{T}_{i=1}^m(\mathcal{R}_i(x, y)) = \mathcal{T}_{i=1}^m\left(\mathcal{I}(A_i(x), B_i(y))\right)$$

$$\widehat{Z} : B'(y) = \widehat{\mathcal{T}}(A'(x), \widehat{\mathcal{R}}(x, y)) = \widehat{\mathcal{T}}\left(A'(x), \mathcal{T}_{i=1}^m(\mathcal{R}_i(x, y))\right)$$

$$B'(y) = \sup_{x \in X} \widehat{\mathcal{T}}\left(A'(x), \mathcal{T}_{i=1}^m\left(\mathcal{I}(A_i(x), B_i(y))\right)\right)$$

- Prologue to **FATI** and **FITA**

$$\widehat{\mathcal{T}}\left(A'(x), \left(\bigcap R_i(x, y)\right)\right) \subseteq \bigcap\left(\widehat{\mathcal{T}}\left(A'(x), R_i(x, y)\right)\right)$$

- In SBR, we have to once again consider the fuzzy IF-THEN rule
- Let the given input be \tilde{x} is A'

Inference in SBR schemes is based on:

- Measure of compatibility/similarity $M(A, A')$
- Modification function J

- Any function $M: \mathcal{F}(X) \times \mathcal{F}(X)$ ¹ $\Rightarrow \langle 0, 1 \rangle$ is said to be a matching function
- A matching function operates on A, A' ² to give a real number in $\langle 0, 1 \rangle$

Interpretation

- Degree of similarity between A and A'
- Subsethood measure of A' in A
- Measure of the compatibility of A' to A

¹ X – non-empty set, $\mathcal{F}(X)$ – fuzzy power set of X

² $A, A' \in \mathcal{F}(X)$

- Zadeh's max-min:

$$M_Z(A, A') = \max_{x \in X} \min(A(x), A'(x))$$

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- Magrez - Smets' measure:

$$M_M(A, A') = \max_{x \in X} \min(N(A(x)), A'(x))$$

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- Measure of subsethood:

$$M_S(A, A') = \min_{x \in X} I(A'(x), A(x))$$

- Zadeh's max-min:

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$$M_S(A, A') = \min_{x \in X} I(A'(x), A(x))$$

- Scalar product:

$$M_C(A, A') = \frac{A \cdot A'}{\max(A \cdot A, A' \cdot A')}$$

- Zadeh's max-min:

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- Magrez - Smets' measure:

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- Measure of subsethood:

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- Scalar product:

$$M_C(A, A') = \frac{A \cdot A'}{\max(A \cdot A, A' \cdot A')}$$

- Disconsistency measure:

$$M_{TK}(A, A') = \sqrt{\frac{\sum_{i=1}^n (A(x_i) - A'(x_i))^2}{n}}$$

- Let Y be a non-empty set and $B \in \mathcal{F}(Y)$
- The modification function J produces a modification $B' \in \mathcal{F}(Y)$ based on s^3 and B^4

J is given by:

$$B'(y) = J(s, B(y)) = J(M(A, A'), B(y)), \quad y \in Y$$

³ $s = M = (A, A')$

⁴ B is the consequence in SBR

- More or Less:

$$J_{ML}(s, B) = B'(x) = \min \left(1, \frac{B(x)}{s} \right)$$

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- Membership Value Reduction:

$$J_{MVR}(s, B) = B'(x) = B(x) \cdot s$$

- In the case of multiple rules in AR we infer the final output by aggregating over the rules
- We use an associative aggregation operation

$$G^5 : \langle 0, 1 \rangle^2 \rightarrow \langle 0, 1 \rangle$$

It is defined by the following:

$$B'(y) = G_{i=1}^m (J(M(A_i, A'), B_i(y))), \quad y \in Y$$

⁵ G is either a t-norm or co-norm

- Let $K: \langle 0, 1 \rangle^2 \rightarrow \langle 0, 1 \rangle$, be an associative and commutative function that combines the matching degrees of A_j to A'_j for all $j = 1, \dots, n$
- K is the combiner in the sequel

The consequence of an individual MISO rule is given by

$$\begin{aligned}B'(y) &= J(K(M(A_1, A'_1), \dots, M(A_n, A'_n)), B(y)), \\&= J(K(s_1, \dots, s_n), B(y)), \quad y \in Y\end{aligned}$$

We infer the final output by aggregating over the rules

$$B'(y) = G_{i=1}^m (J(K_{j=1}^n (M(A_{ij}, A'_j)), B(y))), \quad y \in Y$$

- Some SBR inference schemes along with their inference operations, where T is any t-norm, S is any t-conorm, I is any fuzzy implication and Avg. is the averaging operation

SBR scheme	G	J	K	M
CMI	T	I	T	M_Z
AARS	S_M	J_{MVR}, J_{ML}	Avg.	M_{TK}, M_Z
CDR	T_M	I	-	M_S

$$\frac{\begin{array}{c} \text{if } x \text{ is } A \text{ then } y \text{ is } B \\ x \text{ is } A \end{array}}{y \text{ is } B}$$

Figure: Basic property of GMP (1)

- Rule : if x is A then y is B
- Fact : x is A
- Conclusion : y is B

#basic property

$$B'(y) = \sup_{x \in X'} \mathcal{T}\left(A(x), \mathcal{I}\left(A(x), B(y)\right)\right), y \in Y$$

- Analyzed by when generalized modus ponens coincides with classical modus ponens

$$\frac{\begin{array}{c} \text{if } x \text{ is } A \text{ then } y \text{ is } B \\ x \text{ is } A' \subset A \end{array}}{y \text{ is } B}$$

Figure: **Subset property** of GMP (1)

Fuzzy subset

$$A \subset B \iff ((\forall x \in X : \mu_A(x) \leq \mu_B(x)) \wedge A \neq B)$$

#subset property

$$B''(y) = A'' \circ \mathcal{R}(A, B)$$

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$$B''(y) = \sup_{x \in X} \mathcal{T}\left(A''(x), \mathcal{I}\left(A(x), B(y)\right)\right)$$

$$B''(y) = A'' \circ \mathcal{R}(A, B)$$

$$B''(y) = \sup_{x \in X} \mathcal{T}\left(A''(x), \mathcal{I}\left(A(x), B(y)\right)\right)$$

$$B''(y) \geq \sup_{x \in X} \mathcal{T}\left(A'(x), \mathcal{I}\left(A(x), B(y)\right)\right)$$

$$B''(y) = A'' \circ \mathcal{R}(A, B)$$

$$B''(y) = \sup_{x \in X} \mathcal{T}\left(A''(x), \mathcal{I}\left(A(x), B(y)\right)\right)$$

$$B''(y) \geq \sup_{x \in X} \mathcal{T}\left(A'(x), \mathcal{I}\left(A(x), B(y)\right)\right)$$

$$B''(y) = B'(y)$$

$$\begin{array}{c} \textit{if } x \textit{ is } A \textit{ then } y \textit{ is } B \\ x \textit{ is } \neg A \\ \hline y \textit{ is unknown} \end{array}$$

Figure: **Total indeterminance** of GMP (1)

$\neg A$ represents negation of fuzzy set A by some fuzzy negator

#total indeterminance

$$B'(y) = \sup_{x \in X} \mathcal{T}\left(\bar{A}(x), \mathcal{I}(A(x), B(y))\right)$$

$${}^6X' = \{x \in X \mid \bar{A}(x) = 1\}$$

$$B'(y) = \sup_{x \in X} \mathcal{T}\left(\bar{A}(x), \mathcal{I}(A(x), B(y))\right)$$

$$B'(y) \geq \sup_{x \in X'} \mathcal{T}\left(\bar{A}(x), \mathcal{I}(A(x), B(y))\right)^6$$

$$X' = \{x \in X \mid \bar{A}(x) = 1\}$$

$$B'(y) = \sup_{x \in X} \mathcal{T}\left(\overline{A}(x), \mathcal{I}(A(x), B(y))\right)$$

$$B'(y) \geq \sup_{x \in X'} \mathcal{T}\left(\overline{A}(x), \mathcal{I}(A(x), B(y))\right)^6$$

$$B'(y) = \sup_{x \in X'} \mathcal{T}\left(\overline{A}(x), \mathcal{I}(0, B(y))\right)$$

$${}^6X' = \{x \in X \mid \overline{A}(x) = 1\}$$

$$B'(y) = \sup_{x \in X} \mathcal{T}\left(\bar{A}(x), \mathcal{I}(A(x), B(y))\right)$$

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$$B'(y) = \sup_{x \in X'} \mathcal{T}\left(\bar{A}(x), \mathcal{I}(0, B(y))\right)$$

$$B'(y) = \mathcal{T}(1, 1) = 1$$

$${}^6X' = \{x \in X \mid \bar{A}(x) = 1\}$$

$$\frac{\begin{array}{c} \text{if } x \text{ is } A \text{ then } y \text{ is } B \\ x \text{ is } A' \end{array}}{y \text{ is } B' \supset B}$$

Figure: Superset property of GMP (1)

Fuzzy superset

$$A \supset B \iff ((\forall x \in X : \mu_A(x) \geq \mu_B(x)) \wedge A \neq B)$$

#superset property

We have to consider 2 cases:

- Fuzzy set A' is a fuzzy singleton
- set A' is not a singleton

Consider matching function \mathcal{M} and modification function $\mathcal{J} = \mathcal{I}$, a fuzzy implication (2)

- Superset property

$$\mathcal{I}(p, q) \geq q, \quad p, q \in [0, 1]$$

- Subset property

$$A_1 \subseteq A_2 \implies \mathcal{M}(A_1, A) \geq \mathcal{M}(A_2, A)$$

- Total indeterminance

$$\mathcal{M}(A, \bar{A}) = \mathcal{M}(\bar{A}, A) = 0$$

- Basic property

$$\mathcal{M}(A, A) = 1 \quad \wedge \quad \mathcal{I}(1, y) = y$$

-  Christer, C.; Fuller, R.: Fuzzy Reasoning in Decision Making and Optimization. Physica-Verlag HD, 2012, ISBN 9783790818055, 338 s.
-  Michał Baczyński, B. J.: Fuzzy Implications. Springer, 2008, ISBN 9783540690825, 310 s.

Python notebook - visualization :

https://colab.research.google.com/drive/1IILogFsFL94pFxomR8eR_qon226fNSoS4?usp=sharing