

Semantics for proximity-based logic programming

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Background: unification

Task: find unifiers of terms t to s , i.e., substitutions σ such that $t\sigma = s\sigma$.

Let $a \sim b$ and $b \sim c$. Then

- ▶ $p(b, b)$ not close to $p(a, a)$
- ▶ $p(b, b)$ not close to $p(c, c)$
- ▶ $p(b, b)$ not close to $p(a, c)$

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Task: find (\mathcal{R}, λ) -unifiers of terms t to s , i.e., substitutions σ such that $\mathcal{R}(t\sigma, s\sigma) \geq \lambda$ with cut value λ and similarity relation \mathcal{R} .

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2. from similarity relation to block-based proximity relation

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1. from crisp case to similarity relation
2. from similarity relation to block-based proximity relation
3. from block-based proximity to class-based proximity
4. from minimum T-norm to arbitrary T-norms

Let $a \sim b$ and $b \sim c$. Then

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- ▶ all depending on T-norm and cut value

From unification to logic programming

Literature

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 - ▶ includes declarative and operational semantics

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 - ▶ includes declarative and operational semantics
- ▶ 2023, Julián-Iranzo and Sáenz-Pérez: *“Bousi Prolog - Design and implementation of a proximity-based fuzzy logic programming language”*
 - ▶ only block-based approach
 - ▶ no declarative semantics

Declarative semantics: Interpretation

Theoretical foundation on definite (Horn) clauses $A \leftarrow B_1, \dots, B_n$.

Interpretation function: $\mathcal{I} = \langle \mathcal{D}, \mathcal{V} \rangle$, where

- ▶ \mathcal{D} subset of Herbrand base
- ▶ $\mathcal{V} : \mathcal{D}^n \rightarrow [0, 1]$

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Recursively:

- ▶ $\mathcal{I}(A_1, \dots, A_n) = \bigwedge_{i=1}^n \mathcal{I}(A_i)$
- ▶ $\mathcal{I}(A \leftarrow Q) = \begin{cases} \mathcal{I}(A) & \text{if } \mathcal{I}(A) < \mathcal{I}(Q) \\ 1 & \text{else} \end{cases}$

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Linearization

Needed for defining notions of *model* and *logical consequence*.

Example:

$$\text{lin}(\{p(X, X)\}) = \{p(X, Y) \leftarrow X \sim Y\} = \{p(X, Y) \leftarrow \mathcal{R}(X, Y)\}$$

In proximity case: $\text{lin}(\{p(X, X)\}) =$
 $\{p(Y, Z) \leftarrow Y \sim X, X \sim Z\} = \{p(Y, Z) \leftarrow \mathcal{R}(Y, X) \otimes \mathcal{R}(X, Z)\}$

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Tasks

- ▶ describe fixpoint/immediate consequence operator
- ▶ check *model intersection property*

Operational semantics

Weak (fuzzy) unification

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Weak SLD-resolution

- ▶ decide between computing family of answers or constraints/best answer
- ▶ adapt for linearized programs
- ▶ check for correctness

Further outlook

Rewriting

Builds on *fuzzy matching*:

Find all (\mathcal{R}, λ) -matchers of t to s , i.e., substitutions σ such that $\mathcal{R}(t\sigma, s) \geq \lambda$.

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Quantales

Generalization of “arbitrary T-norms”.