A Fuzzy Logic Programming Environment for Managing Similarity and Truth Degrees*

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FASILL (acronym of “Fuzzy Aggregators and Similarity Into a Logic Language”) is a fuzzy logic programming language with implicit/explicit truth degree annotations, a great variety of connectives and unification by similarity. FASILL integrates and extends features coming from MALP (Multi-Adjoint Logic Programming), a fuzzy logic language with explicitly annotated rules) and Bousi-~Prolog (which uses a weak unification algorithm and is well suited for flexible query answering). Hence, it properly manages similarity and truth degrees in a single framework combining the expressive benefits of both languages. This paper presents the main features and implementations details of FASILL. Along the paper we describe its syntax and operational semantics and we give clues of the implementation of the lattice module and the similarity module, two of the main building blocks of the new programming environment which enriches the FLOPER system developed in our research group.

Keywords: Fuzzy Logic Programming, Similarity Relations, Software Tools

1 Introduction

The challenging research area of Fuzzy Logic Programming is devoted to introduce fuzzy logic concepts into logic programming in order to explicitly treat with uncertainty in a natural way. It has provided a wide variety of PROLOG dialects along the last three decades. Fuzzy logic languages can be classified (among other criteria) regarding the emphasis they assign when fuzzifying the original unification/resolution mechanisms of PROLOG. So, whereas some approaches are able to cope with similarity/proximity relations at unification time [9] [1] [29], other ones extend their operational principles (maintaining syntactic unification) for managing a wide variety of fuzzy connectives and truth degrees on rules/goals beyond the simpler case of true or false [16] [19] [24].

The first line of integration, where the syntactic unification algorithm is extended with the ability of managing similarity/proximity relations, is of special relevance for this work. Similarity/proximity relations put in relation the elements of a set with a certain approximation degree and serve for weakening the notion of equality and, hence, to deal with vague information. With respect to this line, the related work can be summarized as follows:

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Firstly, the pioneering papers [4, 8, 9] and [3], where the concept of unification by similarity was first developed. Note that, we share their objectives, using similarity relations as a basis, but contrary to our proposal, they use the sophisticated (but cumbersome) notions of *clouds*, *systems of clouds* and *closure operators* in the definition of the unification algorithm, that may endanger the efficiency of the derived operational semantics.

More closely tied to our proposal, is the work presented in [29] by Maria Sessa. She defines an extension of the SLD-resolution principle, incorporating a similarity-based unification procedure which is a reformulation of Martelli and Montanari’s unification algorithm [18] where symbols match if they are similar (instead of syntactically equal). The resulting algorithm uses a generalized notion of most general unifier that provides a numeric value, which gives a measure of the approximation degree, and a graded notion of logical consequence. Sessa’s approach to unification can be considered our starting point.

From a practical point of view, similarity-based approaches have produced three main experimental realizations. The first two system prototypes described in the literature were: the fuzzy logic language LIKELOG (*LIKE*ness in *LOG*ic) [2] (an interpreter implemented in PROLOG using rather direct techniques and the aforementioned cloud and closure concepts described in [3, 4, 9, 8]) and SiLog [17] (an interpreter written in Java based on the ideas introduced in [29]). Neither LIKELOG nor SiLog are publicly available, what prevent a real evaluation of these systems, and they seem immature prototypes. In this same line of work, Bousi~Prolog [12, 15], on the other hand, is the first fuzzy logic programming system which is a true PROLOG extension and not a simple interpreter able to execute a weak SLD-resolution procedure. Also it is the first fuzzy logic programming language that proposed the use of proximity relations as a generalization of similarity relations [11]. It is worth saying that, in order to deal with proximity relations, Bousi~Prolog has needed to develop new theoretical [14] and conceptual [28] basis.

A related programming framework, akin to Fuzzy Logic Programming, is *Qualified Logic Programming* (QLP) [26], which is a derivation of van Emden’s *Quantitative Logic Programming* [7] and *Annotated Logic Programming* [16]. In QLP a qualification domain D is associated to a program and their rules annotated with qualification values, resulting a parametric framework: QLP(D). In [5] they introduce similarity relations in their QLP(D) framework by adopting a transformational approach. The new Similarity-based QLP(D) scheme, named SQLP(D), transforms a similarity relation into a set of QLP(D) rules able to emulate a unification by similarity process. In [27, 6] they go a step further integrating constraints and proximity relations in their generic scheme, obtaining a really flexible programming framework named SQCLP.

Ending this section, it is important to say that our research group has been involved both on the development of similarity-based logic programming systems and those that extend the resolution principle, as reveals the design of the Bousi~Prolog language [11, 13, 28], where clauses cohabit with similarity/proximity equations, and the development of the FLOPER system [5] which manages fuzzy programs composed by rules richer than clauses [20, 23]. Our unifying approach is somehow inspired by [6], but in our framework we admit a wider set of connectives inside the body of programs rules. In this paper, we make a first step in our pending task of embedding into FLOPER the *weak unification* algorithm of Bousi~Prolog.

The structure of this paper is as follows. Firstly, in Sections 2 and 3 we formally define and illustrate both the syntax and operational semantics, respectively, of the FASILL language. Next, Section 4 is

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1 Two different programming environments for Bousi~Prolog are available at [http://dectau.uclm.es/bousi/](http://dectau.uclm.es/bousi/).

2 The tool is freely accessible from the Web site [http://dectau.uclm.es/floper/](http://dectau.uclm.es/floper/).
concerned with implementation and practical issues. Finally, in Section 5 we conclude by proposing too further research.

2 The FASILL language

FASILL is a first order language built upon a signature $\Sigma$, that contains the elements of a countably infinite set of variables $\mathcal{V}$, function symbols and predicate symbols with an associated arity – usually expressed as pairs $f/n$ or $p/n$ where $n$ represents its arity –, the implication symbol ($\rightarrow$) and a wide set of others connectives. The language combines the elements of $\Sigma$ as terms, atoms, rules and formulas. A constant $c$ is a function symbol with arity zero. A term is a variable, a constant or a function symbol $f/n$ applied to $n$ terms $t_1, \ldots, t_n$, and is denoted as $f(t_1, \ldots, t_n)$. We allow values of a lattice $L$ as part of the signature $\Sigma$. Therefore, a well-formed formula can be either:

- $r$, if $r \in L$
- $p(t_1, \ldots, t_n)$, if $t_1, \ldots, t_n$ are terms and $p/n$ is an $n$-ary predicate. This formula is called atom. Particularly, atoms containing no variables are called ground atoms, and atoms built from nullary predicates are called propositional variables
- $\xi(\mathcal{F}_1, \ldots, \mathcal{F}_n)$, if $\mathcal{F}_1, \ldots, \mathcal{F}_n$ are well-formed formulas and $\xi$ is an $n$-ary connective with truth function $\xi: L^n \rightarrow L$

**Definition 2.1** (Complete lattice). A complete lattice is a partially ordered set $(L, \leq)$ such that every subset $S$ of $L$ has infimum and supremum elements. Then, it is a bounded lattice, i.e., it has bottom and top elements, denoted by $\bot$ and $\top$, respectively. $L$ is said to be the carrier set of the lattice, and $\leq$ is its ordering relation.

The language is equipped with a set of connectives interpreted on the lattice, including

- aggregators denoted by $\otimes$, whose truth functions $\otimes$ fulfill the boundary condition: $\otimes(\top, \top) = \top$, $\otimes(\bot, \bot) = \bot$, and monotonicity: $(x_1, y_1) \leq (x_2, y_2) \Rightarrow \otimes(x_1, y_1) \leq \otimes(x_2, y_2)$.
- t-norms and t-conorms [25] (also named conjunctions and disjunctions, that we denote by $\&$ and $|$ respectively) whose truth functions fulfill the following properties:
  - Commutative: $\&(x, y) = \&(y, x)$
  - Associative: $\&((x, \&(y, z)) = \&((x, y), z)$
  - Identity element: $\&(x, \top) = x$
  - Monotonicity in each argument: $z \leq t \Rightarrow \{ \&((z, y) \leq \&((t, y) \&((x, z) \leq \&((x, t)$

**Example 1.** In this paper we use the lattice $([0,1], \leq)$, where $\leq$ is the usual ordering relation on real numbers, and three sets of connectives corresponding to the fuzzy logics of Gödel, Łukasiewicz and Product, defined in Figure[7] where labels L, G and P mean respectively Łukasiewicz logic, Gödel logic and product logic (with different capabilities for modeling pessimistic, optimistic and realistic scenarios).

It is possible to include also other connectives. For instance, the arithmetical average, defined by connective $\otimes_{\text{aver}}$ (with truth function $\otimes_{\text{aver}}(x, y) \triangleq \frac{x+y}{2}$), that is a stated, easy to understand connective that does not belong to a known logic. Connectives with arities different from 2 can also be used, like the $\otimes_{\text{very}}$ aggregation, defined by $\otimes_{\text{very}}(x) \triangleq x^2$, that is a unary connective.

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3Here, the connectives are binary operations but we usually generalize them with an arbitrary number of arguments.
Certainly, we are interested in fuzzy binary relations on a syntactic domain. We primarily define similarities on the symbols of a signature, indistinguishable two syntactic symbols which are related by a similarity relation

\[ R \]

**Definition 2.2** (Similarity relation). Given a domain \( \mathcal{U} \) and a lattice \( L \) with a fixed t-norm \( \land \), a similarity relation \( R \) is a fuzzy binary relation on \( \mathcal{U} \), that is a fuzzy subset on \( \mathcal{U} \times \mathcal{U} \) (namely, a mapping \( R : \mathcal{U} \times \mathcal{U} \to L \)), such that fulfills the following properties:\(^4\)

- Reflexive: \( R(x,x) = \top, \forall x \in \mathcal{U} \)
- Symmetric: \( R(x,y) = R(y,x), \forall x,y \in \mathcal{U} \)
- Transitive: \( R(x,z) \geq R(x,y) \land R(y,z), \forall x,y,z \in \mathcal{U} \)

Certainly, we are interested in fuzzy binary relations on a syntactic domain. We primarily define similarities on the symbols of a signature, \( \Sigma \), of a first order language. This makes possible to treat as indistinguishable two syntactic symbols which are related by a similarity relation \( R \). Moreover, a similarity relation \( R \) on the alphabet of a first order language can be extended to terms by structural induction in the usual way \(^{29}\). That is, the extension, \( \hat{R} \), of a similarity relation \( R \) is defined as:

1. let \( x \) be a variable, \( \hat{R}(x,x) = R(x,x) = 1 \),
2. let \( f \) and \( g \) be two \( n \)-ary function symbols and let \( t_1, \ldots, t_n, s_1, \ldots, s_n \) be terms,
   \[
   \hat{R}(f(t_1, \ldots, t_n), g(s_1, \ldots, s_n)) = R(f,g) \land (\land_{i=1}^n \hat{R}(t_i,s_i))
   \]
3. otherwise, the approximation degree of two terms is zero.

Analogously for atomic formulas. Note that, in the sequel, we shall not make a notational distinction between the relation \( R \) and its extension \( \hat{R} \).

**Example 2.** A similarity relation \( R \) on the elements of \( \mathcal{U} = \{ \text{vanguardist}, \text{elegant}, \text{metro}, \text{taxi}, \text{bus} \} \) is defined by the following matrix:

|       | vanguardist | elegant | metro | taxi | bus |
|-------|-------------|---------|-------|------|-----|
| vanguardist | 1          | 0.6     | 0     | 0    | 0   |
| elegant    | 0.6         | 1       | 0     | 0    | 0   |
| metro      | 0           | 0       | 1     | 0.4  | 0.5 |
| taxi       | 0           | 0       | 0.4   | 1    | 0.4 |
| bus        | 0           | 0       | 0.5   | 0.4  | 1   |

It is easy to check that \( R \) fulfills the reflexive, symmetric and transitive properties. Particularly, using the Gödel conjunction as the t-norm \( \land \), we have that:
\[
\hat{R}(\text{taxi}, \text{metro}) \geq \hat{R}(\text{metro}, \text{bus}) \land \hat{R}(\text{bus}, \text{taxi}) = 0.5 \land 0.4.
\]

Furthermore, the extension \( \hat{R} \) of \( R \) determines that the terms \( \text{elegant}(\text{taxi}) \) and \( \text{vanguardist}(\text{metro}) \) are similar, since:
\[
\hat{R}(\text{elegant}(\text{taxi}), \text{vanguardist}(\text{metro})) = R(\text{elegant}, \text{vanguardist}) \land \hat{R}(\text{taxi}, \text{metro}) = 0.6 \land 0.4 = 0.6 \land 0.4 = 0.4.
\]

**Definition 2.3** (Rule). A rule has the form \( A \leftarrow R \), where \( A \) is an atomic formula called head and \( R \), called body, is a well-formed formula (ultimately built from atomic formulas \( B_1, \ldots, B_n \), truth values of

\(^4\)For convenience, \( R(x,y) \), also denoted \( xRy \), refers to both the syntactic expression (that symbolizes that the elements \( x,y \in \mathcal{U} \) are related by \( R \)) and the truth degree \( \mu_{R}(x,y) \), i.e., the affinity degree of the pair \( (x,y) \in \mathcal{U} \times \mathcal{U} \) with the verbal predicate \( R \).

\[ \hat{R}(x,y) \triangleq x \land y \]
\[ \| (x,y) \triangleq x \lor y \]
\[ \| (x,y) \triangleq \max(x,y) \]
\[ \| (x,y) \triangleq \min(x,y) \]
In particular, when the body of a rule is \( r \in L \) (an element of lattice \( L \)), this rule is called fact and can be written as \( A \leftarrow r \) (or simply \( A \) if \( r = \top \)).

**Definition 2.4 (Program).** A program \( \mathcal{P} \) is a tuple \((\Pi, \mathcal{R}, L)\) where \( \Pi \) is a set of rules, \( \mathcal{R} \) is a similarity relation between the elements of \( \Sigma \), and \( L \) is a complete lattice.

**Example 3.** The set of rules \( \Pi \) given below, the similarity relation \( \mathcal{R} \) of Example 2 and lattice \( L = ([0,1], \leq) \) of Example 7 form a program \( \mathcal{P} = (\Pi, \mathcal{R}, L) \).

\[
\Pi = \begin{cases} 
R_1: & \text{vanguardist(hydropolis)} \leftarrow 0.9 \\
R_2: & \text{elegant(ritz)} \leftarrow 0.8 \\
R_3: & \text{close(hydropolis, taxi)} \leftarrow 0.7 \\
R_4: & \text{good_hotel}(x) \leftarrow @\text{aver}(\text{elegant}(x), @\text{very}(\text{close}(x, \text{metro})))
\end{cases}
\]

### 3 Operational Semantics of FASILL

Rules in a FASILL program have the same role than clauses in PROLOG (or MALP [19, 10, 22]) programs, that is, stating that a certain predicate relates some terms (the head) if some conditions (the body) hold.

As a logic language, FASILL inherits the concepts of substitution, unifier and most general unifier (mgu). Some of them are extended to cope with similarities. Concretely, following the line of Bousi–Prolog [11], the most general unifier is replaced by the concept of weak most general unifier (w.m.g.u.) and a weak unification algorithm is introduced to compute it. Roughly speaking, the weak unification algorithm states that two expressions (i.e., terms or atomic formulas) \( f(t_1, \ldots, t_n) \) and \( g(s_1, \ldots, s_n) \) weakly unify if the root symbols \( f \) and \( g \) are close with a certain degree (i.e., \( \mathcal{R}(f, g) = r > \perp \)) and each of their arguments \( t_i \) and \( s_i \) weakly unify. Therefore, there is a weak unifier for two expressions even if the symbols at their roots are not syntactically equals (\( f \not\equiv g \)).

More technically, the weak unification algorithm we are using is a reformulation/extension of the one which appears in [29] for arbitrary complete lattices. We formalize it as a transition system supported by a similarity-based unification relation “\( \Rightarrow \)”. The unification of the expressions \( \mathcal{E}_1 \) and \( \mathcal{E}_2 \) is obtained by a state transformation sequence starting from an initial state \( (G \equiv \{ \mathcal{E}_1 \approx \mathcal{E}_2 \}, \text{id}, \alpha_0) \), where \( \text{id} \) is the identity substitution and \( \alpha_0 = \top \) is the supreme of \( (L, \leq) \): \( \langle G, \text{id}, \alpha_0 \rangle \Rightarrow \langle G_1, \theta_1, \alpha_1 \rangle \Rightarrow \cdots \Rightarrow \langle G_n, \theta_n, \alpha_n \rangle \). When the final state \( (G_n, \theta_n, \alpha_n) \), with \( G_n = \emptyset \), is reached (i.e., the equations in the initial state have been solved), the expressions \( \mathcal{E}_1 \) and \( \mathcal{E}_2 \) are unifiable by similarity with w.m.g.u. \( \theta_n \) and unification degree \( \alpha_n \). Therefore, the final state \( (\emptyset, \theta_n, \alpha_n) \) signals out the unification success. On the other hand, when expressions \( \mathcal{E}_1 \) and \( \mathcal{E}_2 \) are not unifiable, the state transformation sequence ends with failure (i.e., \( G_n = \text{Fail} \)).

The similarity-based unification relation, “\( \Rightarrow \)”, is defined as the smallest relation derived by the following set of transition rules (where \( \forall \text{ar}(t) \) denotes the set of variables of a given term \( t \))

\[
\frac{\langle \{ f(t_1, \ldots, t_n) \approx g(s_1, \ldots, s_n) \} \cup E, \theta, r_1 \rangle \quad \mathcal{R}(f, g) = r_2 > \perp}{\langle \{ t_1 \approx s_1, \ldots, t_n \approx s_n \} \cup E, \theta, r_1 \land r_2 \rangle}
\]

---

In order to subsume the syntactic conventions of MALP, in our programs we also admit weighted rules with shape “\( A \leftarrow i \ R \) with \( \nu \)”, which are internally treated as “\( A \leftarrow (\nu \& i; \mathcal{R}) \)” (this transformation preserves the meaning of rules as proved in [21]).
Rule 1 decomposes two expressions and annotates the relation between the function (or predicate) symbols at their root. The second rule eliminates spurious information and the fourth rule interchanges the position of the symbols to be handled by other rules. The third and fifth rules perform an occur check of variable X in a term t. In case of success, it generates a substitution \{X/t\}; otherwise the algorithm ends with failure. It can also end with failure if the relation between function (or predicate) symbols in \( R \) is \( \bot \), as stated by Rule 6.

Usually, given two expressions \( E_1 \) and \( E_2 \), if there is a successful transition sequence, \( \{E_1 \approx E_2\}, id, T \) \( \Rightarrow^* \{\emptyset, \theta, r\} \), then we write that \( \text{wmgu}(E_1, E_2) = \{\theta, r\} \), being \( \theta \) the weak most general unifier of \( E_1 \) and \( E_2 \), and \( r \) their unification degree.

Finally note that, in general, a w.m.g.u. of two expressions \( E_1 \) and \( E_2 \) is not unique \([29]\). Certainly, the weak unification algorithm only computes a representative of a w.m.g.u. class, in the sense that, if \( \theta = \{x_1/t_1, \ldots, x_n/t_n\} \) is a w.m.g.u., with degree \( \beta \), then, by definition, any substitution \( \theta' = \{x_1/s_1, \ldots, x_n/s_n\} \), satisfying \( R(s_i/t_i) > \bot \), for any \( 1 \leq i \leq n \), is also a w.m.g.u. with approximation degree \( \beta' = \beta \land (\bigwedge_{i=1}^{n} R(s_i/t_i)) \), where “\( \land \)” is a selected t-norm. However, observe that, the w.m.g.u. representative computed by the weak unification algorithm is one with an approximation degree equal or greater than any other w.m.g.u. As in the case of the classical syntactic unification algorithm, our algorithm always terminates returning a success or a failure.

Next, we illustrate the weak unification process in the following example.

**Example 4.** Consider the lattice \( L = ([0,1], \leq) \) of Example 7 and the relation \( R \) of Example 2. Given terms \( \text{elegant}(\text{taxi}) \) and \( \text{vanguardist}(\text{metro}) \), it is possible the following weak unification process:

\[
\langle \{\text{elegant}(\text{taxi}) \approx \text{vanguardist}(\text{metro})\}, id, 1 \rangle \overset{1}{\Rightarrow} \langle \{\text{taxi} \approx \text{metro}\}, id, 0.6 \rangle \overset{1}{\Rightarrow} \langle \{\}, id, 0.4 \rangle
\]

Also it is possible the unification of the terms \( \text{elegant}(\text{taxi}) \) and \( \text{vanguardist}(X) \), since:

\[
\langle \{\text{elegant}(\text{taxi}) \approx \text{vanguardist}(X)\}, id, 1 \rangle \overset{4}{\Rightarrow} \langle \{\text{taxi} \approx X\}, id, 0.6 \rangle \overset{4}{\Rightarrow} \langle \{\}, \{X/\text{taxi}\}, 0.6 \rangle
\]

and the substitution \( \{X/\text{taxi}\} \) is their w.m.g.u. with unification degree 0.6.

In order to describe the procedural semantics of the FASILL language, in the following we denote by \( \mathcal{C}[A] \) a formula where \( A \) is a sub-expression (usually an atom) which occurs in the -possibly empty- context \( \mathcal{C}[\] \) whereas \( \mathcal{C}[A/A'] \) means the replacement of \( A \) by \( A' \) in the context \( \mathcal{C}[\] \). Moreover, \( \mathcal{V} ar(s) \) denotes the set of distinct variables occurring in the syntactic object \( s \) and \( \theta[\mathcal{V} ar(s)] \) refers to the substitution obtained from \( \theta \) by restricting its domain to \( \mathcal{V} ar(s) \). In the next definition, we always consider that \( A \) is the selected atom in a goal \( \mathcal{G} \) and \( L \) is the complete lattice associated to \( \Pi \).
Definition 3.1 (Computational Step). Let $Q$ be a goal and let $\sigma$ be a substitution. The pair $\langle Q; \sigma \rangle$ is a state. Given a program $\langle \Pi, R, L \rangle$ and a t-norm $\land$ in $L$, a computation is formalized as a state transition system, whose transition relation $\xrightarrow{\cdot}$ is the smallest relation satisfying these rules:

1) Successful step (denoted as $SS$): 

\[
\frac{\langle Q[A], \sigma \rangle}{\langle Q[A/A \land \theta] \theta, \sigma \theta \rangle}
\]

2) Failure step (denoted as $FS$): 

\[
\frac{\langle Q[A], \sigma \rangle}{\langle Q[A/A \land \theta] \theta, \sigma \theta \rangle}
\]

3) Interprete step (denoted as $IS$): 

\[
\frac{\langle Q[A/A \land \theta] \theta, \sigma \theta \rangle}{\langle Q[A/A \land \theta] \theta, \sigma \theta \rangle}
\]

A derivation is a sequence of arbitrary length $\langle Q; \sigma \rangle \xrightarrow{\cdot} \langle Q'; \sigma \rangle$. As usual, rules are renamed apart.

Example 5. Let $\mathcal{P} = \langle \Pi, R, L \rangle$ be the program from Example 3 and $Q = \text{good\_hotel}(X)$ be a goal. It is possible to perform these two derivations for $\mathcal{P}$ and $Q$: 

Figure 2: Screen-shot of a work session with FLOPER managing a FASILL program
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4 Implementation of FASILL in FLOPER

During the last years we have developed the FLOPER tool, initially intended for manipulating MALP programs. In its current development state, FLOPER has been equipped with new features in order to cope with more expressive languages and, in particular, with FASILL. The new version of FLOPER is freely accessible in the URL http://dectau.uclm.es/floper/?q=sim where it is possible to test/download the new prototype incorporating the management of similarity relations. In this section we briefly describe the main features of this tool before presenting the novelties introduced in this work.

FLOPER has been implemented in Sicstus Prolog v.3.12.5 (rounding about 1000 lines of code, where our last update supposes approximately a 30% of the final code) and it has been recently equipped with a graphical interface written in Java (circa 2000 lines of code). More detailed, the FLOPER system consists in a JAR (Java archive) file that runs the graphical interface. This JAR file calls a PROLOG file containing the two main independent blocks: 1) the Parsing block parses FASILL files into two kinds of PROLOG code (a high level platform-independent PROLOG program and a set of facts to be used by FLOPER), and 2) the Procedural block performs the evaluation of a goal against the program, implementing the procedural semantics previously described. This code is completed with a configuration file indicating the location of the PROLOG interpreter as well as some other data.

FLOPER provides a traditional command interpreter. When the command interpreter is executed, it offers a menu with a set of commands grouped in four submenus:

- “Program Menu”: includes options for parsing a FASILL program from a file with extension “.fp3”, saving the generated PROLOG code to a “.pl” file, loading/parsing a pure PROLOG program, listing the rules of the parsed program and cleaning the database.

6 The MALP language is nowadays fully subsumed by the new FASILL language just introduced in this paper, since, given a FASILL program $\mathcal{P} = (\Sigma, \mathcal{R}, L)$, if $\mathcal{R}$ is the identity relation (that is, the one where each element of a signature $\Sigma$ is only similar to itself, with the maximum similarity degree) and $L$ is a complete lattice also containing adjoint pairs [19], then $\mathcal{P}$ is a MALP program too.
“Lattice Menu”: allows the user to change and show the lattice (implemented in PROLOG) associated to a fuzzy program through options lat and show, respectively.

“Similarity Menu”: option sim allows the user to load a similarity file (with extension “.sim”, and whose syntax is detailed further in the Similarity Module subsection) and tnorm sets the conjunction to be used in the transitive closure of the relation.

“Goal Menu”: by choosing option intro the user introduces the goal to be evaluated. Option tree draws the execution tree for that goal whereas leaves only shows the fuzzy computed answer contained on it, and depth is used for fixing its maximum depth.

The syntax of FASILL presented in Section I is easily translated to be written by a computer. As usual in logic languages, variables are written as identifiers beginning by an upper case character or an underscore “_”, while function and predicate symbols are expressed with identifiers beginning by a lower case character, and numbers are literals. Terms and atoms have the usual syntax (the function or predicate symbol, if no nullary, is followed by its arguments between parentheses and separated by a
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Connectives are labeled with their name immediately after. The implication symbol is written as “<-”, and each rule ends with a dot. Additionally it is possible to include pure PROLOG expressions inside the body of a rule by encapsulating them between curly brackets “{ }”, and PROLOG clauses between the dollar symbol “$”, together with FASILL rules.

The graphical interface (written in Java) supports a friendly interaction with the user, as seen in Figure 3. The graphical interface shows three areas. The leftmost one draws the project tree (grouping each category of file into its own directory). In the right part, the upper area displays the selected file of the tree and the lower one shows the code and the solutions of executing a goal. This interface groups files into projects which include a set of fuzzy files (.fpl), PROLOG files (.pl), similarity files (.sim), script files - containing a list of commands to be executed consecutively - (.vfs) and just one lattice file (.lat). When executing a goal, the tool considers the whole program merged from the set of files, thus obtaining only one fuzzy program, one similarity relation, one lattice and one PROLOG file.

The lattice module. Lattices are described in a .lat file using a language that is a subset of PROLOG where the definition of some predicates are mandatory, and the definition of aggregations follows a certain syntax. The mandatory predicates are member/1, that identifies the elements of the lattice, bot/1 and top/1, that stand for the infimum and supremum elements of the lattice, and leq/2, that implements the ordering relation. Predicate members/1, that returns in a list all the elements of the lattice, is only required if it is finite. Connectives are defined as predicates whose meaning is given by a number of clauses. The name of a predicate has the form and_label, or_label or agr_label depending on whether it implements a conjunction, a disjunction or an aggregator, where label is an identifier of that particular connective (this way one can define several conjunctions, disjunctions and other kind of aggregators instead of only one). The arity of the predicate is n + 1, where n is the arity of the connective that it implements, so its last parameter is a variable to be unified with the truth value resulting of its evaluation.

\[ \text{AGR}_{\text{LABEL}}(r_1, \ldots, r_n, R). \]
\[ R = r. \]

Example 6. For instance, the following clauses show the PROLOG program modeling the lattice of the real interval \([0, 1]\) with the usual ordering relation and connectives (conjunction and disjunction of the Product logic, as well as the average aggregator):

member(X) :- number(X), 0=<X, X=<1.
and_prod(X,Y,Z) :- Z is X*Y.
bot(0).
or_prod(X,Y,Z) :- U1 is X*Y, U2 is X+Y, Z is U2-U1.
top(1).
agr_aver(X,Y,Z) :- U1 is X*Y, U2 is X+Y, Z is U2-U1.

The similarity module. We describe now the main novelty introduced in the tool, that is the ability to take into account a similarity relation. The similarity relation \( \mathcal{R} \) is loaded from a file with extension .sim through option sim. The relation is represented following a concrete syntax:

\[ \mathcal{R} ::= \langle \text{Sim} \rangle \langle \text{Relation} \rangle \mid \langle \text{Sim} \rangle \langle \text{Id}_f \rangle [ \langle \text{Id}_g \rangle ] \sim [ \langle \text{Id}_n \rangle ] \sim \langle \text{r} \rangle \sim \langle \text{tnorm} \rangle \sim \langle \text{norm} \rangle \]

The Sim option parses expressions like “\( f \sim g = r \)”, where \( f \) and \( g \) are propositional variables or constants and \( r \) is an element of \( L \). It also copes with expressions including arities, like “\( f/n \sim g/n = r \)” (then, \( f \) and \( g \) are function or predicate symbols). In this case, both arities have to be the same. It is
also possible to explicit, through a line like “\( \sim \text{tnorm} = \langle \text{label} \rangle \)” the conjunction to be used further in the construction of the transitive closure of the relation. Internally FLOPER stores each relation as a fact \( r \) in an ad hoc module \( \text{sim} \) as \( r(f/n,g/n,r) \), where \( n = 0 \) if it has not been specified (that is, the symbol is considered as a constant). The \( \text{sim} \) file contains only a small set of similarity equations that FLOPER completes by performing the reflexive, symmetric and transitive closure. The first one simply consists of the assertion of the fact \( r(A,A,\top) \). The symmetric closure produces, for each \( r(a,b,r) \), the assertion of its symmetric entry \( r(b,a,r) \) if there is not already some \( r(b,a,r') \) where \( r \leq r' \) (in this case \( r(a,b,r) \) will be rewritten as \( r(a,b,r') \) when considering \( r(b,a,r) \)). The transitive closure is computed by the next algorithm\(^7\), where \( \wedge \) stands for the conjunction specified by the directive “\( \text{tnorm} \)” and “\( \text{assert} \)” and “\( \text{retract} \)” are self-explainable and defined as in PROLOG:

\[
\text{Transitive Closure}
\begin{align*}
\text{forall } r(A,B,r_1) & \text{ in sim} \\
\text{forall } r(B,C,r_2) & \text{ in sim} \\
& r = r_1 \wedge r_2 \\
& \text{if } r(A,C,r') \text{ in sim and } r' < r \\
& \quad \text{retract } r(A,C,r') \text{ from sim} \\
& \quad \text{retract } r(C,A,r') \text{ from sim} \\
& \text{end if} \\
& \text{if } r(A,C,r') \not\text{ in sim} \\
& \quad \text{assert } r(A,C,r) \text{ in sim} \\
& \quad \text{assert } r(C,A,r) \text{ in sim} \\
& \text{end if} \\
& \text{end forall}
\end{align*}
\]

It is important to note that, it is not relevant if the user provides (apparently) inconsistent similarity equations, since FLOPER automatically changes the user values by the appropriate approximation degrees in order to preserve the properties of a similarity. For instance, if a user provides a set of equations such as, \( a \sim b = 0.8 \), \( b \sim c = 0.6 \) and \( a \sim c = 0.3 \), after the application of our algorithm for the construction of a similarity, results in the set of equations \( a \sim b = 0.8 \), \( b \sim c = 0.6 \) and \( a \sim c = 0.6 \), which positively preserves the transitive property\(^8\).

**Example 7.** Let \( L \) be the lattice \( ([0,1], \leq) \). To illustrate the enhanced expressiveness of FASILL, consider the program \( \langle \Pi, R, L \rangle \) that models the concept of good hotel, that is, an elegant hotel that is very close to a metro entrance, as seen in Figure\(^3\). Here, we use an average aggregator defined as \( @_{\text{avg}}(x, y) \triangleq (x + y)/2 \), whereas very is a linguistic modifier implemented as well as an aggregator (with arity 2) with truth function \( @_{\text{very}} x \triangleq x^2 \). The similarity relation \( R \) states that elegant is similar to vanguardist, and metro to bus and (by transitivity) to taxi:

\[
\begin{align*}
\sim \text{tnorm} &= \text{godel} \\
\text{elegant/1} \sim \text{vanguardist/1} &= 0.6 \\
\text{metro} \sim \text{bus} &= 0.5 \\
\text{bus} \sim \text{taxi} &= 0.4
\end{align*}
\]

We also state that the \( t \)-norm to be used in the transitive closure is the conjunction of G\"odel (i.e., the infimum between two elements). For this program (the set of rules of Figure\(^3\) the lattice \( L \) and the similarity relation, \( R \), just described before), the goal \( \text{good\_hotel}(X) \) produces two fuzzy computed answers: \( <0.4, \{ X/\text{ritz}\} > \) and \( <0.38, \{ X/\text{hydropolis}\} > \). Each one corresponds to the leaves of

\(^7\) It is important to note that this algorithm must be executed right after performing the symmetric, reflexive closure.

\(^8\) For simplicity, we have omitted the equations obtained during the construction of the reflexive, symmetric closure.
the tree depicted in Figure 3. Note that for reaching these solutions, a failure step was performed in the derivation of the left-most branch, whereas in the right-most one (and this is the crucial novelty w.r.t. previous versions of the FLOPER tool) there exist two successful steps exploiting the similarity relation atom close(hydropolis, metro), which illustrates the flexibility of our system.

Ending this section, it is worthy to say that our approach differs from the one presented in [6] since they employ a combination of transformation techniques to first extract the definition of a predicate “~”, simulating weak unification in terms of a set of complex program rules that extends the original program. Finally, this predicate “~” is reduced to a built-in proximity/similarity unification operator (in this case not implemented by rules and very close to the implementation of our weak unification algorithm) that highly improves the efficiency of their previous programming systems.

4.1 FLOPER online

Aside from the textual and graphical interfaces seen above, we have recently developed a web page aiming at offering the FLOPER system via the Internet, without requiring any further installation. The interaction with the system is possible through the URL http://dectau.uclm.es/floper/?q=sim/test. Under the title of FLOPER Online it shows an interface divided in two areas. The Input area, shown in Figure 4, is located in the upper part of the window, and the Output area is in the lower part of the window, and is illustrated by Figure 5.

The Input area shows three boxes. The first one, under the label “FASILL program” is intended to contain a set of FASILL rules, that is, the fuzzy program. The second one contains the lattice associated to the previously introduced program. By default it includes the ([0, 1], ≤) lattice, obviously expressed as a PROLOG program following the restrictions previously detailed, with the usual operators. The user is free to implement here any complete lattice as far as it fulfils the syntactic constrains. In the third box the user can write a set of similarity equations using the program’s signature. After these boxes the user can introduce a goal in a text box (in Figure 4, the goal is good_hotel(X)). Finally, by clicking the Submit button, the fuzzy program, together with its associated lattice and the similarity relation, is sent to the server with the goal to be executed.

The result appears in the Output area in two ways. In the first place, under the label F.c.a’s for goal… (including the proper goal), the system shows the fuzzy computed answers for the introduced program and goal. In the figure this corresponds to < 0.4, {X/Ritz} > and < 0.38, {X/hydropolis} >, as expected. Further, in the box below, the derivation tree is depicted in a textual way.

The tool has been implemented as a php page inside the web of FASILL. This php document sends the content of the text boxes (the FASILL program, the lattice, the similarity equations and the goal) to itself via the “post” method. When the php loads again with non empty “post” parameters, it creates files in the server to host the FASILL program, the lattice and the similarity equations, and calls the PROLOG interpreter. Then, it consults the FLOPER environment, loads the files and queries the goal. The output of this task (that is, the corresponding f.c.a’s and execution tree) are finally shown in the window.

Each state contains its corresponding goal and substitution components and they are drawn inside yellow ovals. Computational steps, colored in blue, are labeled with the program rule they exploit in the case of successful steps or the annotation “R0” in the case of failure steps (observe that, “R0” is a simple notation and do not correspond with any existing rule). Finally, the blue circles annotated with the word “is”, correspond to interpretive steps.
5 Conclusions and Future Work

This paper describes an extension of the FLOPER system to cope with the twofold integrated fuzzy programming language FASILL, whose procedural principle is centered upon a weak—instead of a syntactic—unification algorithm based on similarity relations. After a brief introduction of the syntactic and operational features of FASILL, we describe the implementation details of the renewed FLOPER system which gives support to FASILL. We center our attention on the description of the similarity module, providing insights of the internal representation of a similarity relation and its automatic construction, via built-in closure algorithms. Also we describe the new tool of FLOPER online, that allows the execution of FASILL programs thru the web.

On the other hand, in [11, 10, 22] we provided some advances in the design of declarative semantics and/or correctness properties regarding the development of fuzzy logic languages dealing with similarity/proximity relations (Bousi~Prolog) or highly expressive lattices modeling truth degrees (MALP). As a matter of future work we want to establish that analogous—but reinforced—formal properties also hold in the language FASILL.
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Figure 5: Screenshot of the FLOPER online tool output

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