# Logic Programming with Functions over Order-Sorted Feature Terms 

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#### Abstract

LIFE is an experimental programming language proposing to integrate logic programming, functional programming, and object-oriented programming. It replaces first-order terms with $\psi$-terms, data strucures which allow computing with partial information. These are approximation structures denoting sets of values. LIFE further enriches the expressiveness of $\psi$-lerms with functional dependency constraints. Whereas LIFE's relations defined as Horn-clauses use $\psi$-term unification for parameter-passing, LIFE's functions use $\psi$-term matching (i.e., one-way unification). We explain the meaning and use of functions in LFE declaratively as solving partial information constraints. These constraints do not attempt to generate their solutions but behave as demons filtering out anything else. In this manner, LIFE functions act as declarative coroutines.


The paradox of culture is that language [...] is too tinear, not comprehensive enough, too slow, too limited, too constrained, too unnatural, too much * product of its own evolution, and too anificial. This means that [man] must constantly keep in mind the limitions language places upon him.

Edward T. Hall, BeyondCulture.

## 1 Introduction

This paper is an informal, albeit precise and detailed, overview of the operational mechanism underlying functional reduction in the context of a logic programming framework. It is formulated using order-sorted feature terms as data structures scen as constraints. This mechanism and data structure are used in the language LIFE [5].

### 1.1 The task

LIFE extends the computational paradigm of Logic Programming in two essential ways:

- by using a data structure richer than that provided by first-order constructor terms; and,
- by allowing interpretable functional expressions as bona fide terms.

The first extension is based on $\psi$-terms which are attributed partially-ordered sorts denoting sets of objects [1, 2]. In particular, $\psi$-terms gencralize first-order constructor terms in their rolce as data structures in that they are endowed with a unification operation denoting type intersection. This gives an clegant means to incorporate a calculus of multiple inheritance into symbolic programming. Importantly, the denotation-as-value of constructor terms is
replaced by the denotation-as-approximation of $\psi$-terms. A consequence of this is that the notion of fully defined element, or ground term, is no longer available. Hence, such familiar tools as variable substitutions, instantiation, unification, etc., must be reformulated in the new setting [5].

The second extension deals with building into the unification operation a means to reduce functional expressions using definitions of interpretable symbols over data patterns. ${ }^{1}$ Our basic idea is that unification is no longer seen as an atomic operation by the resolution rulc. Indecd, since unification amounts to normalizing a conjunction of equations, and since this normalization process commutes with resolution, these equations may be left in a normal form that is not a fully solved form. In particular, if an equation involves a functional expression whose arguments are not sufficiently instantiated to match a definiens of the function in question, it is simply left untouched. Resolution may proceed until the arguments are proven to match a definition from the accumulated constraints in the context [3]. This simple idea turns out invaluable in practice. Here are a few benefits.

- Such non-declarative heresies as the is predicate in Prolog and the freeze meta-predicate in some of its extensions $[14 ; 10]$ are not needed.
- Functional computations are determinate and do not incur the overhead of the search stralcgy needed by logic programming.
- Higher-order functions are casy to return or pass as arguments since functional variables can be bound to partially applicd functions.
- Functions can be called before the arguments are known, frecing the programmer from having to know what the data dependencies are.
- It provides a powerful scarch-space pruning facility by changing "generate-and-test" scarch into demon-controlled "test-and-gencrate" search.
- Communication with the external world is made simple and clean [9].
- More generally, it allows concurrent computation. Synchronization is obtained by checking entailment $[13,15]$.

There are two orthogonal dimensions to clucidate regarding the use of functions in LIFE:

- characterizing functions as approximation-driven coroutines; and,
- constructing a higher-order model of LIFE approximation structures.

This present article is concerned only with the first item, and therefore considers the case of first-order rules defining partial functions over $\psi$-terms.

### 1.2 The method

The most direct way to explain the issuc is with an example. In LIFE, one can define functions as usual; say:

$$
\begin{array}{ll}
\operatorname{fact}(0) & \rightarrow 1 \\
\operatorname{fact}(N: \operatorname{int}) & \rightarrow N * \operatorname{fact}(N-1) .
\end{array}
$$

[^0]More intercsting is the possibility to compute with partial information. For example:

$$
\begin{aligned}
& \text { minus }(\text { negint }) \rightarrow \text { posint } . \\
& \text { minus }(\text { posint }) \rightarrow \text { negint. } \\
& \text { minus }(\text { zero }) \rightarrow \text { zero. } .
\end{aligned}
$$

Let us assume that the symbols int, posint, negint, and zero have been defined as sorts with the approximation ordering such that posint, zero, negint are pairwise incompatible subsorts of the sort int (i.e., posint $\wedge$ zero $=1$, negint $\wedge$ zero $=\perp_{\text {, posint } \wedge \text { negint }}=\perp$ ). This is declared in LIFE as int $:=\{$ posint; zero; negint $\}$. Furthcrmore, we assume the sort definition posint $:=$ \{posodd; poseven\}; i.e., posodd and poseven are subsorts of posint and mutually incompatible.

The LIFE qucry $Y=$ minus $(X$ : poseven)? will return $Y=$ negint. The sort poseven of the actual parameter is incompatible with the sort negint of the formal parameter of the first rule defining the function minus. Thercfore, that rule is skipped. The sort poseven is more specific than the sort posint of the formal parameter of the second rule. Hence, that rule is applicable and yields the result $Y=n e g i n t$.

The LIFE query $\gamma=$ minus( $X$ : string) will fail. Indeed, the sort string is incompatible with the sort of the formal parameter of every rule defining minus.

Thus, in order to determine which of the rules, if any, defining the function in a given functional expression will be applied, two tests are necessary:

- verify whether the actuall parameter is more specific than or equal to the formal parameter;
- verify whether the actual parameter is at all compatible with the formal parameter.

What happens if both of these tests fail? Consider the query consisting of the conjunction:

$$
Y=\operatorname{minus}(X: \operatorname{int}), X=\operatorname{minus}(z e r o) ?
$$

for example. Like Prolog, LIFE follows a left-to-right resolution strategy and examines the equation $Y=$ minus $(X: i n t)$ first. However, both foregoing tests fail and deciding which rule to use among those defining minus is inconclusive. Indecd, the sort int of the actual parameter in that call is neither more specific than, nor incompatible with, the sort negint of the first rule's formal parameter. Therefore, the function call will residuate on the variable $X$. This means that the functional evaluation is suspended pending more information on $X$. The second goal in the query is treated next. There, it is found that the actual parameter is incompatible with the first two rules and is the same as the last rule's. This allows reduction and binds $X$ to zero. At this point, $X$ has been instantiated and therefore the residual equation pending on $X$ cam be reexamined. Again, as before, a redex is found for the last rule and yields $Y=$ zero.

The two tests above can in fact be worded in a more general setting. Viewing data structures as constraints, "more specific" is simply a particular case of constraint entailment. We will say that a constraint disentails another whenever their conjunction is unsatisfiable; or, equivalently, whenever it entails its negation. In particular, first-order matching is deciding entailment between constraints consisting of equations over first-order terms. Similarly, deciding unifiability of first-order terms amounts to deciding "compatibility" in the sense used informally above.

The suspension/resumption mechanism illustrated in our example is repeated each time a residuated actual parameter becomes more instantiated from the context; i.e., solving
other parts of the query. Therefore, it is most beneficial for a practical algorithm testing entailment and disentailment to be incremental. This means that, upon resumption, the test for the instantiated actual parameter builds upon partial results obtained by the previous test. One outcome of the results presented in this paper is that it is possible to build such a test; namely, an algorithm deciding simultancously two problems in an incremental manner-entailment and disentailment. The technique that we have devised to do that is called relative simplification of constraints.

This technique is relevant in the general framework of concurrent constraint logic programming, represented by, e.g., the guarded Horn-clause scheme of Maher [13], Concurrent Constraint Programming (CCP) [15], and Kcrnel Andorra Prolog (KAP) [11]. These schemes are parametcrized with respect to an abstract class of constraint systems. An incremental test for entailment and disentailment between constraints is needed for advanced control mechanisms such as delaying, coroutining, synchronization, committed choice, and deep constraint propagation. LIFE is formally an instance of this scheme, namely a CLP language using a constraint system based on order-sorted feature structures [6]. It employs a related, but limited, suspension strategy to enforce deterministic functional application. Roughly, these systems are concurrent thanks to a new effective discipline for procedure parameter-passing that we could describe as "call-by-constraint-entailment" (as opposed to Prolog's call-by-unification).

In this paper, we recall the basic terminology and notation of LIFE, unification and matching, and we sketch the essence of relative simplification. Formal material rewording everything in rigorous terms can be found in [4]. Nevertheless, in the final section we include the formal version of the simplification rules specifying the algorithms discussed informally in this paper, and state the theoretical results proven in [4].

## 2 LIFE Data Structures

The data objects of LIFE are $\psi$-terms. They are structures built out of sorts and features. $\psi$ Terms are partially ordered as data descriptions to reflect more specific information content. A $\psi$-term is said to match another one if it is a more specific description. For first-order terms, a matching substitution is a variable binding which makes the more general term equal to the more specific one. This notion is not appropriate here. Unification is introduced as taking the greatest lower bound (GLB) with respect to this ordering.

### 2.1 Sorts and features

Sorts are symbols. They are mcant to denote sets of values. Here are a few examples: person, int, true, $3.5, \perp, T$. Note that a value is assimilated to a singleton sort. We call $\mathcal{S}$ the set of all sorts. They come with a partial ordering $\leq$, meant to reflect set inclusion. ${ }^{2}$ For example,

- $\perp \leq$ john $\leq$ man $\leq$ person $\leq T$;
- $\perp \leq$ truc $\leq$ bool $\leq T$;
$\bullet \perp \leq 2 \leq$ poseven $\leq i n t \leq T$.

[^1]

Fig. 1. A partial order of sorts

The sorts $T$ (top) and $\perp$ (bottom) are respectively the greatest and the least sort in $\mathcal{S}$ and denote respectively the whole domain of interpretation and the empty set.

Sorts also come with a GLB operation $\wedge$. For example,

- person $\wedge$ male $=$ man;
- male $\wedge$ female $=$ hermaphrodite;
- man $\wedge$ woman $=\perp$;
etc., which can be visualized as shown in Figure 1. We will refer back to this figure in several examples to come.

Features (or autribute labels) are also symbols and used to build $\psi$-terms by attaching attributes to sorts. The set of feature symbols is called $\mathcal{F}$. We will use words and natural numbers as features. The latter are handy to specify attributes by positions as subterms in first-order terms. Examples of feature symbols are age, spouse, 1, 2.

## $2.2 \psi$-Terms

Basic $\psi$-terms are the simplest form of $\psi$-terms. They arc:

- variables; e.g., $X, \gamma, Z, \ldots$
- sorts; e.g., person, int, true, 3.5, T, ...
- tagged sorts; c.g., $X: T, Y:$ person, ...

Stand-alone variables are always implicitly sorted by $T$, and stand-alone sorts are always implicitly tagged by some variable occurring nowhere else. Thus, one might say that a basic $\psi$-term is always of the form variable : sort.

Fcatures are used to build up more complex $\psi$-terms. Thus, the following $\psi$-term is obtained from the $\psi$-term person by attaching the feature age typed by the $\psi$-term int: ${ }^{3}$

$$
x: \text { person }(\text { age } \Rightarrow 1: \text { int }) .
$$

The sort at the root of a $\psi$-term, here person, is called its principal sort. A $\psi$-term can be seen as a record structure. Features correspond to field identifiers, and fields are, in turn, associated to $\psi$-terms. These are flexible records in the sense that variably many fields may be altached to the principal sort. For example, we can augment the $\psi$-term above with another feature:

$$
\begin{aligned}
& X: \text { person(age } \Rightarrow I: \text { int }, \\
& \quad \text { spouse } \Rightarrow Y: \text { person(age } \Rightarrow J: \text { int })) .
\end{aligned}
$$

This $\psi$-term denotes the set of all objects $X$ of sort person (in the intended domain), whose value $I$ under the function age is of sort int, whose value $Y$ under the function spouse is of sort person, and the value. $J$ of $Y$ under the function age is of sort int.

The following $\psi$-term is more specific, in the sense that the above set becomes smaller if one further requires that the values $I$ and $J$ coincide; namely, age $(X)=\operatorname{age}($ spouse $(X)$ ):

$$
\begin{aligned}
& x: \text { person(age } \Rightarrow I: \text { int, } \\
& \text { spouse } \Rightarrow Y: \operatorname{person}(\text { age } \Rightarrow I)) .
\end{aligned}
$$

It denotes the subset of individuals in the previous set of person's whose age is the same as their spouse's. This $\psi$-term uses a coreference thanks to sharing the variable $/$. The next $\psi$-term is even more specific, since it contains an additional (circular) coreference; namely, $X=\operatorname{spouse}($ spouse $(X))$ :

$$
\begin{aligned}
& X: \text { person(age } \Rightarrow I: \text { int, } \\
& \text { spouse } \Rightarrow Y: \text { person (age } \Rightarrow I, \\
& \\
& \text { spouse } \Rightarrow X)) .
\end{aligned}
$$

It denotes the set of all individuals in the previous set whose spouse's spouse is the individual in question. Note that only variables that are used as coreference tags need to be put explicilly; i.e., those that occur at least twice.

To be well-formed, the syntax of a $\psi$-term requires three conditions to be satisfied: (1) the sort $\perp$ may not occur; (2) at most one occurrence of each variable has a sort; (3) all the features allached to a sort are pairwise different. These conditions are necessary to ensure that a $\psi$-term expresses coherent information. For example, $X:$ man(friend $\Rightarrow X:$ woman , violating Condition (2), is not a $\psi$-term, bul $X: \operatorname{man}(f r i e n d \Rightarrow X)$ is.

As for ordering, a $\psi$-term is made more specific through:

- sort refinement; e.g., $X:$ int $\leq U: T$;
- adding fcalures typed by $\psi$-tcrms; e.g., $X: \mathrm{T}$ (age $\Rightarrow$ int $) \leq U: \mathrm{T}$;
- adding coreference; e.g., $X: T($ likes $\Rightarrow X) \leq U: T($ likes $\Rightarrow V)$.

Note that, as record structures, $\psi$-terms are both record types and record instances. They, in addition, permit mixing type and value information. Finally, they also permit constraining records with equations on their parts.

[^2]

Fig. 2. An OSF-Graph

## $2.3 \psi$-Terms as graphs

There is a straightforward representation of a $\psi$-term as a rooted directed graph. Let us assume that every variable is explicilly sorted (if necessary, by the sort $T$ ) and every sort is explicilly tagged (if necessary, by a single-occurrence variable). The nodes of the graph are the variables, their labels are the corrcsponding sorts; for every feature mapping one variable $X$ to another one $Y$ there is an are $(X, Y)$ labeled by that feature. One node is marked as the root (whose label is called the root sort or the principal sort of the $\psi$-term).

For example, the $\psi$-term:

$$
\begin{gathered}
X_{1}: \text { person }\left(\text { name } \Rightarrow X_{2}: \text { id (first } \Rightarrow X_{3}: \text { string, },\right. \\
\text { last } \left.\Rightarrow X_{4}: \text { string }\right) \\
\text { spouse } \Rightarrow X_{5}: \text { person }\left(\text { name } \Rightarrow X_{6}: \text { id }\left(\text { last } \Rightarrow X_{4}\right),\right. \\
\text { spouse } \left.\left.\Rightarrow X_{1}\right)\right) .
\end{gathered}
$$

corresponds to the OSF-graph shown in Figure 2.

## $2.4 \boldsymbol{\psi}$-Terms as values

One particular interpretation is readily available for $\psi$-terms. Namely, the syntactic interpretation $\Psi$ whose domain is the set of all $\psi$-cerms. Note that $\psi$-terms have a dual personality. They are syntactic objects (graphs) representing the values of the domain of $\Psi$, and they also are types which denote sets. In the particular case of the interpretation $\Psi$, they denote subsets of the domain of $\Psi$; i.e., sets of $\psi$-terms. We shall see this dual view does not lead to paradox, au contraire.

In the interpretation $\Psi$, a sort $s \in S$ denotes the set of all $\psi$-terms whose root sort is a subsort of $s$. A feature $\ell \in \mathcal{F}$ denotes the function mapping a $\boldsymbol{\psi}$-term to its sub- $\boldsymbol{\psi}$-term under that feature, or to $T$, if there is none.

Thus, a sort denotes the set of all $\psi$-term values which, as $\boldsymbol{\psi}$-term types, are more specific than the basic $\psi$-term $s$. In fact, it is possible to show that in general a $\psi$-term denotes the sel of all $\psi$-terms which are more specific than the $\psi$-term itself. This is the " $\psi$-tcrms as filters" principle cstablished in [5]. It yields directly the fact that the partial ordering $\leq$ on $\psi$-terms is cxactly set-inclusion of the sets denoted by the $\psi$-terms in the $\psi$-term domain.

### 2.5 Feature trees as values

We obtain two other examples of OSF-algebras when we "compress" the $\psi$-term domain by identifying valucs. In a first step, we say that two $\psi$-terms which are equal up to variable renaming represent the same value of the domain, or: two isomorphic graphs are identified. We call the OSF-algebra hereby obtained $\Psi_{0}$.

It is well known that a rooted directed graph represents a unique rational tree obtained by unfolding. Hence, unfolding an OSF-graph yields what we call a feature tree. Such a tree is one whose nodes are labeled with sorts and whose edges are labeled with features. Therefore, we can also identify $\psi$-terms which represent the same rational tree. The domain hereby obtained is essentially the feature tree structure $\mathcal{T}$ introduced first in [7] and [8].

### 2.6 Unification of' $\psi$-terms

We say that $\psi_{1}$ is unifiable with $\psi_{2}$ if $\psi_{1} \wedge \psi_{2} \neq \perp$; i.e., if there exist $\psi$-terms with non-cmpty denotations which are more specific than both $\psi_{1}$ and $\psi_{2}$. One can show that then there exists a unique (up to variable renaming) $\psi$-term $\psi$ which is the most general of all these, the 'greatest lower bound' (GLB) of $\psi_{1}$ and $\psi_{2}$, written $\psi=\psi_{1} \wedge \psi_{2}$.

For the sel denotation of $\psi$-terms, $\wedge$ is exactly set intersection. An important result illustrating the significance of the $\psi$-term interpretation $\Psi$ is that $\psi_{1}$ is unifiable with $\psi_{2}$ if and only if the intersection of the two scts denoted by $\psi_{1}$ and $\psi_{2}$ in the $\psi$-term domain is non-emply.

### 2.7 Constraints and $\psi$-terms

We also view a $\psi$-term logically as a constraint formula by flattening it into what we call its dissolved form. For case of notation, we shall write $(X: \psi)$ to indicate that the root variable of the $\psi$-term $\psi$ is $X$.

More preciscly, the $\psi$-tcrm $X: s\left(\ell_{1} \Rightarrow\left(X_{1}: \psi_{1}\right), \ldots, \ell_{n} \Rightarrow\left(X_{n}: \psi_{n}\right)\right)$ corresponds to the conjunction of the constraint $X: s \& X . \ell_{1} \doteq X_{1} \& X . \ell_{n} \doteq X_{n}$ and of the constraints corresponding to $\psi_{1}, \ldots, \psi_{n}$. A basic $\psi$-term $X: s$ corresponds to the sort constraint $X: s$. For example, the $\psi$-tcrm:

$$
\begin{gathered}
\psi \equiv X: \text { person }(\text { likes } \Rightarrow X, \\
\text { age } \Rightarrow Y: i n t)
\end{gathered}
$$

is identified with the constraint:

$$
\psi \equiv X: \text { person \& } X . a g e \doteq \gamma \& \gamma: \text { int \& X.likes } \doteq X
$$

Thus, the constraint $\psi$ is a conjunction of atomic sort constraints of the form $X: s$ and atomic feature constraints of the form $X . \ell \doteq \gamma$. The interpretation of the sort and feature constraints over the intended domain is straightforward, given that sorts are interpreted as subsets of the domain and features as unary functions over the domain.

A value lies in the set denoted by the $\psi$-term $\psi$ in an interpretation $\mathcal{I}$ if and only if the constraint $X \doteq Z \& \psi$ is satisfiable in the interpretation $\mathcal{I}$, with that value assigned to the variable $X$, and $Z$ being the root variable of $\psi$. All variables of $\psi$ are implicitly existentially quantified. This reflects our view of $\psi$-terms as sel-denoting types.

### 2.8 Rules for Unification

Unifying ( $X_{1}: \psi_{1}$ ) and ( $X_{2}: \psi_{2}$ ) amounts to deciding satisfiability of the conjunction $\psi_{1} \& \psi_{2} \& X_{1} \doteq X_{2}$. Thus, the unification algorithm can be specified in terms of constraint normalization rules. A constraint containing the conjunction over the line is rewritten into an equivalent constraint by replacing this conjunction by the constraint under the line. We only need four rules that are illustrated schematically on an example below. (Refer to the sorts of Figure 1.)

## Equality:

$$
\frac{\ldots X \text { : person \& } U \text { :male \& } U \doteq X \ldots}{\ldots . . \text { person \& } X: \text { male \& } U \doteq X \ldots}
$$

Sorts:

$$
\frac{\ldots X: \text { person \& } X: \text { male } \ldots}{\ldots X: \operatorname{man} \ldots}
$$

Featurcs:

$$
\frac{\ldots \text { X.likes } \doteq Y \& \quad \text { X.likes } \doteq V \ldots}{\ldots \text { X.likes } \doteq Y \& V \doteq Y \ldots}
$$

Clash:

$$
\frac{\ldots x: \perp \ldots}{\perp}
$$

One can show that a constramt is satisfiable if and only if it is normalized to a constraint different from the false constraint $\perp$. If we identify every constraint containing a sort constraint of the form $X: \perp$ with the false constraint, we omit the the clash rule.

In particular, the $\psi$-terms $\left(X_{1}: \psi_{1}\right)$ and $\left(X_{2}: \psi_{2}\right)$ are unifiable if and only if $\psi_{1} \& \psi_{2} \& X_{1} \doteq X_{2}$ is normalized into a constraint $\psi$ different from $\perp$. This constraint $\psi$ corresponds, apart from its equalities (between variables), to the $\psi$-term (unique up to variable renaming) $\psi_{1} \wedge \psi_{2}$.

## 3 Relative Simplification

We use the framework of first-order logic to transform the combined entailment/disentailment problem into one that can be solved by the relative simplification algorithm.

### 3.1 Matching and entailment

In the remainder of this paper, when considering the matching problem $\psi_{1} \leq \psi_{2}$, we will refer to $\psi_{1}$ as the actual parameter and its variables (named $X, Y, Z, \ldots$ ) as global, and to $\psi_{2}$ as the formal parameter and its variables (named $U, V, W, \ldots$ ) as local.

In the Concurrent Constraint Logic Programming framework, the matching problem gencralizes to the entailment problem; namely, whether the actual constraint, also called context, entails the formal constraint, also called guard [13, 15].

First observe that, for example, the first-order term $t_{1}=f(Z, f(Y, Y))$ matches the term $t_{2}=f(W, V)$, and that the implication:
$\forall X \forall \gamma \forall Z(X \doteq f(Z, f(Y, Y)) \rightarrow \exists U \exists V \exists W(X \doteq U \& U \doteq f(W, V)))$
is valid. Gencrally, the term $t_{1}$ matches $t_{2}$ (noted $t_{1} \leq t_{2}$ ) if and only the implication $X \doteq t_{1} \rightarrow \exists U \exists V\left(X \doteq U \& U \doteq t_{2}\right)$ is valid, where $V$ stands for all variables of $t_{2}$. More shorlly, $X \doteq t_{1}$ entails $X \doteq U \& U \doteq t_{2}$.

Note, however, that there is an cssential difference between $\psi$-term matching and firstorder term matching. For example, the term $f(a, a)$ matches the term $f(V, V)$. This is true because first order terms denote individuals. This is no longer true in LIFE. For example, the $\psi$-term $X: \int(1 \Rightarrow Y: i n i, 2 \Rightarrow Z: i n t)$ does not match the $\psi$-term $U: f(1 \Rightarrow V, 2 \Rightarrow V)$ as two occurrences of the same sort does not entail that the individuals in that sort be equal. Therefore, $X: s\left(1 \Rightarrow \psi_{1}, 2 \Rightarrow \psi_{2}\right)$ is less specific than the $\psi$-term $U: s(1 \Rightarrow V, 2 \Rightarrow V)$ only if the root variables of $\psi_{1}$ and $\psi_{2}$ are identical (or bound together).

This does not mean that values and operations on them are not available in LIFE. ${ }^{4}$ What the above point illustracs is that to recognize that a sort is a fully determined value, and hence to enforce identity of all its distinct occurrences, one needs this information declared explicilly, in effect adding an axiom to the formalization of such sorts. So-declared extensional sorts can then be treated accordingly thanks to an additional inference rule (being a minimal non-bottom sort is not sufficient). Without this rule, however, equality of distinct occurrences cannot be entailed and the behavior illustrated is the only correct one. The point of this paper being independent if this issue, we shall omit this additional rule.

The fact that $\left(X: \psi_{1}\right) \leq\left(U: \psi_{2}\right)$, i.e., the $\psi$-term $\left(X: \psi_{1}\right)$ matches the $\psi$-term ( $U: \psi_{2}$ ), translates into the fact that the corresponding constraint $\psi_{1}$ entails the constraint $\psi_{2} \& U \doteq X$. This means that the implication $\psi_{1} \rightarrow \exists U, V, W \ldots \psi_{2} \& U \doteq X$ is valid. Here, $\exists U, V, W \ldots$ indicates that all local variables are existentially quantified. The global variables are universally quantified.

### 3.2 Entailment of general constraints

We will now give a precise explanation of a fact which is well-known for constructor terms. An actual parameter $t_{1}$ matches a formal parameter $t_{2}$ if and only if the unification of the two terms binds only variables of $\iota_{2}$, but no variable of $t_{1}$. In other words, only local, but no global, variables are instantiated.

[^3]The unification of the term $t_{1}=f(Z, f(Y, Y))$ and the term $t_{2}=f(W, V)$ yields the variable bindings $W \doteq Z$ and $V \doteq f(Y, Y)$. On the other hand, the conjunction:

$$
X \doteq f(Z, f(Y, Y)) \& U \doteq X \& U \doteq f(W, V)
$$

is equivalent to:

$$
X \doteq f(Z, f(Y, \gamma)) \&(U \doteq X \& V \doteq f(\gamma, \gamma) \& W \doteq Z)
$$

and the last part of this conjunction is valid if the local variables $U, V, W$ are existentially quantificd.

This is the general principle which underlies the relative simplification algorithm. Namely, the actual constraint $\psi_{1}$ entails $\psi_{2} \& U \doteq X$ if and only if the following holds. Their conjunction $\psi_{1} \& \psi_{2} \& U \doteq X$ is equivalent to the conjunction $\psi_{1} \& \psi_{2}^{\prime}$ of the actual constraint $\psi_{1}$ and a constraint $\psi_{2}^{\prime}$ which is valid if existentially quantified over the local variables. In our case, $\psi_{2}^{\prime}$ will be a conjunction of equalities binding local to global variables. Formally,

$$
\vDash \psi_{1} \rightarrow \exists U, V, W, \ldots \psi_{2} \& U \doteq X
$$

if and only if there exists a formula $\psi_{2}^{\prime}$ such that

$$
\vDash\left(\psi_{1} \& \psi_{2} \& U \doteq X\right) \leftrightarrow\left(\psi_{1} \& \psi_{2}^{\prime}\right) \text { and } \vDash \exists U, V, W \ldots \psi_{2}^{\prime} .
$$

This statement is correct since validity of the implication $\psi_{1} \rightarrow \exists U \psi_{2} \& U \doteq X$ is the same as the validity of the equivalence $\left(\psi_{1} \&\left(\exists U \psi_{2} \& U \doteq X\right)\right) \leftrightarrow \psi_{1}$. This fact is analogous to the fact that a set is the subset of another one if and only if it is equal to the intersection of the two. The condition $\vDash \exists U, V, W \ldots \psi_{2}^{\prime}$ in the statement expresses that $\psi_{1} \&\left(\exists U, V, W, \ldots \psi_{2}^{\prime}\right)$ is cquivalent to $\psi_{1}$.

### 3.3 Towards relative simplification

Operationally, in order to show that $\left(X: \psi_{1}\right) \leq\left(U: \psi_{2}\right)$ holds, it is sufficient to show that the conjunction $\psi_{1} \& \psi_{2} \& U \doteq X$ is cquivalent to $\psi_{1} \& \psi_{2}^{\prime}$, where $\psi_{2}^{\prime}$ is some constraint which, existentially quantified over the variables of $\psi_{2}$, is valid. In our case, again, $\psi_{2}^{\prime}$ will be a conjunction of cqualitics binding variables of $\psi_{2}$ to variables of $\psi_{1}$.

Therefore, in order totest $\left(X: \psi_{1}\right) \leq\left(U: \psi_{2}\right)$, we willapply successively the unification rules on the constraint $\psi_{1} \& \psi_{2} \& U \doteq X$ if they do not modify $\psi_{1}$. We obtain three kinds of transformations which are illustrated schematically below. (Refer to the sorts of Figure 1.)

Equality:

$$
\frac{\ldots X \doteq Y}{\ldots} \ddagger
$$

Sorts:

$$
\frac{\ldots X: \operatorname{man} \& U \doteq X \& U: \text { person } \ldots}{\ldots X: \operatorname{man} \& U \doteq X \ldots}
$$

Features:

$$
\frac{\ldots X . l i k e s \doteq Y \text { \& } U \doteq X \& ~ \& ~ U . l i k e s \doteq V \ldots}{\ldots X . l i k e s \doteq Y \text { \& } U \doteq X \& V \doteq Y \ldots}
$$

The equality rule is derived from the corresponding unification rule, which has to be restricted to modify only the formal constraint. If the actual constraint contains an equality between two global variables, then occurrences of one of them may be climinated for the other. A global variable is never climinated for a local one.

The sort rule corresponds to two applications of unification rules, first the climination of the local by the global variable, and then the reduction of two sort constraints on the same variable (here $X: \operatorname{man} \& X:$ person) to one sort constraint (namely $X: m a n \wedge$ person). Clearly, if the "global sort" is a subsort of the "local sort" then this application does not modify the global constraint. The feature rule works quite similarly.

For example, the rules above can be used to show that the $\psi$-term:

$$
\psi_{1} \equiv X: \operatorname{man}(\text { likes } \Rightarrow Y: \text { perwon, age } \Rightarrow /: \text { int })
$$

matches the $\psi$-icrm:

$$
\psi_{2} \equiv U: \text { person }(\text { likes } \Rightarrow V)
$$

Namely, the constraint $\psi_{1} \& \psi_{2} \& U \doteq X:$

$$
\begin{aligned}
& \quad X: \text { man \& X.likes } \doteq Y \& Y: p e r s o n \& X . a g e \doteq I \& I: i n t \\
& \& U: p e r s o n \& U . l i k e s \doteq V \\
& \& U \doteq X
\end{aligned}
$$

is normalized into:

$$
\begin{aligned}
& X: \text { man \& } X . l i k e s \doteq Y \& Y: p e r s o n \& X . a g e \doteq I \& I: \text { int } \\
& \& V \doteq Y \text { \& } U \doteq X ;
\end{aligned}
$$

that is,

$$
\psi_{1} \& V \doteq Y \& U \doteq X
$$

Clcarly, $\exists U \exists V(V \doteq Y \& U \doteq X)$ is valid. Therefore, the constraint $\psi_{1}$ entails the constraint $\psi_{2} \& U \doteq X$.

### 3.4 Relative simplification for entailment

The rules above are such that $\psi_{1} \& \psi$ rewrites to $\psi_{1} \& \psi^{\prime} ;$ i.e., the global constraint $\psi_{1}$ is not modificd by the simplification. In this case, we say that the constraint $\psi$ simplifies to $\psi^{\prime}$ relatively to the actual constraint $\psi_{1}$. In other words, $\psi_{1}$ acts as a context relatively to which simplification of $\psi$ is carried out. In general, this context formula may be any formula. Hence, we can reformulate the rules above as relative-simplification rules. We use the notation $\frac{\phi}{\psi^{\prime}}|\phi|$ to mean that $\psi$ is simplified into $\psi^{\prime}$ relatively to the context formula $\phi$. Schematically,

Equality:

$$
\left.\frac{\ldots U \doteq X \ldots}{\ldots U \doteq Y \ldots} \quad \right\rvert\, \ldots x \doteq \gamma \ldots 1
$$

Sorts:

$$
\left.\frac{\ldots U \doteq X \& U: \text { person } \ldots}{\ldots U \doteq X \& \ldots} \right\rvert\, \ldots x: \operatorname{man} \ldots 1
$$

Features:

$$
\left.\frac{\ldots U \doteq X \& U . l i k e s \doteq V \ldots}{\ldots U \doteq X \& V \doteq Y \ldots} \right\rvert\, \ldots \text {. } \ddagger \text { likes } \doteq r \ldots \mid
$$

Using these rules, the constraint $\psi_{2} \equiv U \doteq X \& U:$ person \& U.likes $\doteq V$ in the previous example simplifics to $\psi_{2}^{\prime} \equiv U \doteq X \& V \doteq Y$ relatively to:

$$
\psi_{1} \equiv X: m a n \& X . l i k e s \doteq Y \& Y: \text { person } \& X . a g c \doteq I \& I: \text { int }
$$

Invariance of relative simplification is the following property. If $\psi$ simplifies to $\psi^{\prime}$ relatively to $\phi$, then the conjunction of $\psi$ with $\phi$ is equivalent to the conjunction of $\psi^{\prime}$ with $\phi$.

This invariance justifies the correctness of the relative simplification algorithm with respect to entailment. Namely, if $\psi$ simplifies to $\psi^{\prime}$ relatively to $\phi$ and $\psi^{\prime}$ consists only of equations binding local variables, then $\phi$ entails $\psi$.

Proof of completeness of the algorithm needs the assumption that the set $\mathcal{F}$ of features is infinite. Note that exacily thanks to the infiniteness of $\mathcal{F}$ our framework accounts for flexible records; i.e., the indefinite capacity of adding ficlds to records.

### 3.5 Relative simplification for disentailment

If the result of the matching test $\psi_{1} \leq \psi_{2}$ is negative, i.e., the actual constraint does not entail the formal constraint, then we must know more; namely, whether the two terms are non-unifiable. Non-unifiability is equivalent to the fact that the actual parameter will not match the formal one even when further instantiated; e.g., when further constraints are attached as conjuncts. Logically, this amounts to saying that a context formula $\phi$ disentails a guard constraint $\psi$ if and only if the conjunction $\phi \& \psi$ is unsatisfiable. In terms of relative simplification, $\phi$ disentails $\psi$ if and only if $\dot{\psi}$ simplifies to the false constraint $\perp$ relatively to $\phi$.

For cxample, $X$ : male is non-unifiable with $U$ : woman. ${ }^{5}$ The constraint $U$ : woman \& $U \doteq X$ simplifies to $\perp$ relatively to the constraint $X:$ male, since woman $\wedge$ male $=$ $\perp$, using a rule of the form indicated below, and then the Clash rule.

Sorts:

$$
\left.\left.\frac{\ldots U \doteq X \& U: \text { woman } \ldots}{\ldots U \doteq X \& U: \text { woman } \wedge \text { male } \ldots} \right\rvert\, \ldots x: \text { male } \ldots\right\}
$$

The following example shows that a sort clash cannot always be detected by comparing sorts in the formal constraint one by one with sorts in the actual constraint; i.e., one needs several steps with intermediate sort intersections.

The $\psi$-term $Z: T($ likes $\Rightarrow X:$ male, friend $\Rightarrow Y:$ female $)$ is non-unifiable with the $\psi$ term $W: T($ likes $\Rightarrow U$ : person, friend $\Rightarrow U)$. The constraint $\phi \equiv X$ : male \& $Y$ : female disentails the constraint $\psi \equiv U \doteq X \& U \doteq \gamma \& U$ : person. Operationally, the constraint $\psi$ simplifies to $\perp$ relatively to the context $\phi$. Here are the steps needed to determine this:

[^4]```
    \(\ldots U \doteq X \& U \doteq Y \& U:\) person..
\(\ldots U \doteq X \& U \doteq Y \& U\) : person \(\wedge\) male..
    \(\ldots U \doteq X \& U \doteq Y \& U: \operatorname{man} \wedge\) fomale...
    \(\perp\)
```

There is an issue regarding the enforcing of functionality of features in the simplification of a constraint $\psi$ relatively to a context $\phi$. This may be explained as follows. Let us suppose that two global variables $X$ and $Y$ become bound to the same local variable $U$. Then,

- the context $\phi$ entails the constraint $\psi$ only if $\phi$ contains $X \doteq Y$; and,
- the context $\phi$ disentails the constraint $\psi$ if the same path of features starting from $X$ and $Y$, respectively, leads to variables $X^{\prime}$ and $Y^{\prime}$, respectively, whose sorts are incompatible.

There are essentially two cases, depending on whether a new local variable has to be introduced or not. Each case is illustrated in the next two examples.

The $\psi$-term: ${ }^{6}$

$$
\begin{array}{r}
\phi \equiv Z: \top\left(\text { likes } \Rightarrow X: \top\left(\text { age } \Rightarrow I_{1}: \text { poseven }\right),\right. \\
\\
\text { friend } \left.\Rightarrow \gamma: \top\left(\text { age } \Rightarrow I_{2}: \text { posodd }\right)\right)
\end{array}
$$

is non-unifiable with the $\psi$-term:

$$
\begin{array}{r}
\psi \equiv W: \mathrm{T}(\text { likes } \Rightarrow U, \\
\\
\text { friend } \Rightarrow U)
\end{array}
$$

That is, the constraint $\phi$ disentails the constraint $\psi$. Operationally, with the context $\phi$, the constraint $\psi$ simplifies, in a first step, to:

$$
W \doteq Z \& U \doteq X \& U \doteq Y
$$

Then, using the rule:

$$
\frac{\ldots U \doteq X \& U \doteq Y \ldots}{\ldots U \doteq X \& U \doteq Y \& J \doteq I_{1} \& J \doteq I_{2} \ldots} \quad\left\{\ldots X . a g e \doteq I_{1} \& Y . a g e \doteq I_{2} \ldots\right\}
$$

where $J$ is a new variable, to:

$$
W \doteq Z \& U \doteq X \& U \doteq Y \& J \doteq I_{1} \& J \doteq I_{2}
$$

and finally to $\perp$, since the sorts of $I_{1}$ and $I_{2}$ (poseven and posodd) are incompatible.
The rules enforce the following property: a global variable is never bound to more than one local variable. Therefore, if the variable $X$ or the variable $Y$ is already bound to a local variable, no new local variable must be introduced. This is illustrated by the second example.

The $\psi$-term:

$$
\begin{gathered}
\phi \equiv Z: T\left(\text { likes } \Rightarrow X: T\left(\text { age } \Rightarrow I_{1}: \text { poseven }\right),\right. \\
\\
\text { friend } \Rightarrow Y: T\left(\text { age } \Rightarrow I_{2}: \text { posodd }\right), \\
\\
\text { age } \left.\Rightarrow I_{1}\right)
\end{gathered}
$$

[^5]is non-unifiable with the $\psi$-term:
\[

$$
\begin{aligned}
\psi \equiv W: T & (\text { likes } \Rightarrow U \\
& \text { friend } \Rightarrow U(\text { age } \Rightarrow J) \\
& \text { age } \Rightarrow J)
\end{aligned}
$$
\]

Operationally, with the context $\phi$, the constraint $\psi$ simplifies, in a first step, to:

$$
W \doteq Z \& U \doteq X \& U \doteq Y \& J \doteq I_{1}
$$

Then, using the rule:

$$
\left.\frac{\ldots U \doteq X \& U \doteq Y \& J \doteq I_{1} \ldots}{\ldots U \doteq X \& U \doteq Y \& J \doteq I_{1} \& J \doteq I_{2} \ldots} \quad 1 \ldots X . a g e \dot{ } \doteq I_{1} \& Y . a g e \doteq I_{2} \ldots\right]
$$

where $J$ is a new variable, to:

$$
W \doteq Z \& U \doteq X \& U \doteq Y \& J \doteq I_{1} \& J \doteq I_{2}
$$

and finally to $\perp$, for the same reason as above.
In order to be complete with respect to disentailment, the algorithm must keep track of all pairs of variables $(X, Y), \ldots,\left(X^{\prime}, Y^{\prime}\right)$ whose cquality is induced by the binding of $X$ and $Y$ to the same local variable. That is, it must propagate cqualities along features. In our presentation, it will be conceptually sufficient to refer explicitly to the actual equalities binding the global variables to a common local variable. Practically, this can of course be done more efficiently.

### 3.6 Specifying the relative simplification algorithm

If $\psi \& U \doteq X$ simplifies to $\psi^{\prime}$ relatively to $\phi$ and no relative-simplification rule can be applied further, then:

- $\phi$ entails $\psi \& U \doteq X$; formally,

$$
\vDash \phi \rightarrow \exists U, V, W \ldots(\psi \& U \pm X)
$$

if and only if $\psi^{\prime}$, with the variables of $\psi$ existentially quantified, is valid; formally:

$$
\vDash \exists U, V, W \ldots \psi^{\prime}
$$

- $\phi$ disentails $\psi \& U \doteq X$; formally:

$$
\vDash \phi \rightarrow \neg \exists U, V, W \ldots(\psi \& U \doteq X)
$$

if and only if $\psi^{\prime}=\perp$.
This test is incremental. Namely, every relative simplification of the constraint $\psi$ to some constraint $\psi^{\prime}$ relatively to the context $\phi$ is also a relative simplification relatively to an incremented context $\phi \& \phi^{\prime}$, for any constraint $\phi^{\prime}$.

Recapitulating, our original goal was a simultancous test of matching and non-unifiability for two given $\psi$-terms $\psi_{1}$ and $\psi_{2}$. This test was recast as a test of entailment and disentailment for the constraints to which the $\psi$-terms dissolve. Namely, if $X$ and $U$ are the root variables of $\psi_{1}$ and $\psi_{2}$, respectively, the test whether $\psi_{1}$ entails or disentails $\psi_{2} \& U \doteq X$.

In our setting, the entailment test succeeds if and only if $\psi_{2}^{\prime}$ is a conjunction of matching equations; i.e., of the form $\psi_{2}^{\prime} \equiv U \doteq X \& V \doteq Y \& W \doteq Z \ldots$, where the local variables $U, V, W, \ldots$ are all different.

Feature Decomposition:
(13.1)

$$
\frac{\psi \& U . \ell \doteq V \& U . \ell \doteq W}{\psi \& U . \ell \doteq V \& W \doteq V}
$$

## Sort Intersection:

$$
\begin{equation*}
\frac{\psi \& U: s \& U: s^{\prime}}{\psi \& U: s \wedge s^{\prime}} \tag{13.2}
\end{equation*}
$$

Variable Elimination:
(B.3) $\frac{\psi \& U \doteq V}{\psi[V / U] \& U \doteq V}$
if $U \in \operatorname{Var}(\psi)$ and $U \neq V$

Inconsistent Sort:
(13.4) $\frac{\psi \& X: \perp}{\perp}$

Variable Clean-up:
(13.5) $\frac{\psi \& U \doteq U}{\psi}$

Fig. 3. Basic Simplification

## 4 Simplification Rules

We give now the formal version of the simplification rules specifying the algorithms discussed informally in this paper, and state the theoretical results proven in [4]. Here, OSF-constraints are conjunctions of atomic sort constraints of the form $X: s$, atomic fcature constraints of the form $X . \ell \doteq Y$, and equations $X \doteq Y$. Their interpretations over OSF-algebras arc straightforward (cf., Scction 2.7).

### 4.1 Unification-Satisfiability

The first algorithm, called basic simplification, determines whether a constraint $\phi$ is consistent; $i . e .$, if it is satisfiable in some OSF-algebra $\mathcal{A}$-and, thercfore, in particular in $\Psi$. Unification of two $\psi$-terms reduces to this problem.

Given an OSF-constraint $\phi$, it can be normalized by choosing non-deterministically and applying any applicable rule among the transformations rules shown in Figure 3 until none applies. A rule transforms the numerator into the denominator. The expression $\phi[X / Y]$ stands for the formula obtained from $\phi$ after replacing all occurrences of $Y$ by $X$.

Feature Decomposition:
(F.1) $\frac{\psi \& U . \ell \doteq V \& U . \ell \doteq W}{\psi \& U . \ell \doteq V \& \dot{W} \doteq V}$

Relative Feature Decomposition:
(F.2) $\frac{\psi \& U \doteq X \& U . \ell \doteq V}{\psi \& U \doteq X \& V \doteq Y}$

Relative Feature Equality:

$$
\begin{array}{ll}
\psi \& U \doteq X_{1} \& U \doteq X_{2} \& V \doteq Y_{1} & \text { if } X_{1}, \ell \doteq Y_{1} \in \phi, X_{2}, \ell \doteq Y_{2} \in \phi  \tag{F.3}\\
\psi \& U \doteq X_{1} \& U \doteq X_{2} \& V \doteq Y_{1} \& V \doteq Y_{2} & \text { and } V \doteq Y_{2} \notin \psi
\end{array}
$$

Variable Introduction:
(F.4) $\frac{\psi \& U \doteq X_{1} \& U \doteq X_{2}}{\psi \& U \doteq X_{1} \& U \doteq X_{2} \& V \doteq Y_{1} \& V \doteq Y_{2}}$
if $X_{1}, \ell \doteq Y_{1} \in \phi, X_{2} . \ell \doteq Y_{2} \in \phi$ and $Y_{1} \notin \operatorname{Var}(\psi)$ and $Y_{2} \notin \operatorname{Var}(\psi)$ where $V$ is a new variable

Fig. 4. Simplification relatively to $\phi$ : Fcatures

The rules of Figure 3 are solution-preserving (i.e., equivalence transformations), finite terminating, and confluent (modulo variable renaming).

The effectuality of the basic-simplification system is summed up in the following statement:

Effectuality of basic-simplification The constraint $\psi$ is satisfiable if and only if the normal form $\psi^{\prime}$ of $\psi$ is different from the false constraint, i.e., $\psi^{\prime} \neq \perp$.

### 4.2 Matching-Entailment/Disentailment

The nextalgorithm, called relative simplification, determines whether a consistent constraint $\phi$ entails or disentails a constraint $\psi$ which is the dissolved form of some $\psi$-term with root variable $U$. That is, it decides two problems simultancously:

- the validity of the implication $\forall \mathcal{X}(\phi \rightarrow \exists \mathcal{U} .(\psi \& U \doteq X))$;
- the unsatisfiability of the conjunction $\phi \& \psi \& U \doteq X$.

Matching of two $\psi$-terms reduces to the first of the two problems, non-unifiability to the second.

The relative-simplification system for OSF-constraints is given by the rules in Figures 4, 5, and 6. An OSF-constraint $\psi$ simplifics to $\psi^{\prime}$ relatively to $\phi$ by a simplification rule $\rho$ if $\frac{\psi}{\psi^{\prime}}$ is an instance of $\rho$ and the applicability condition (on $\phi$ and on $\psi$ ) is satisficd. We say that $\psi$ simplifies to $\psi^{\prime}$ relatively to $\phi$ if it does so in a finite number of steps.

Relative Variable Elimination:


## Equation Entailment:

(E.2) $\frac{\psi \& U \doteq X \& U \doteq Y}{\psi \& U \doteq X}$

$$
\text { if } X=Y \text { or if } X \doteq Y \in \phi
$$

Fig. 5. Simplification relatively to $\phi$ : Equations

The relative-simplification system preserves an important invariant property: the conjunction $\phi \& \psi$ is equivalent to the conjunction $\phi \& \psi^{\prime}$. Again, the rules are finite terminating, and conlluent (modulo variable renaming).

A sct of bindings $U_{i} \doteq X_{i}, i=1, \ldots, n$ is a functional binding if all the variables $U_{i}$ are mutually distinct.

The effectuality of the relative-simplification system is summed up in the following statement:

> Effectuality of relative-simplification The solved OSF-constraint $\phi$ entails (resp., disentails) the OSF-constraint $\exists U .(U \doteq X \& \psi)$ if and only if the normal form $\psi^{\prime}$ of $\psi \& U \doteq X$ relatively to $\phi$ is a conjunction of equations making up a functional binding (resp., is the false constraint $\psi^{\prime}=\perp$ ).

## 5 Conclusion

We have presented informally the essence of LIFE that is relevant to functions and explained the gist of our method for deciding incrementally entailment and disentailment of constraints. We call this new technique relative simplification. Reading this paper will provide an claborate intuition of the technical details reported formally in [4]. There, we expound the concept of relative simplification as a general proof-theoretic method for proving guards in concurrent constraint logic languages using guarded rules. The specific relative simplification rules for OSF-constraints are given and proven correct and complete. Then, residuation is naturally explained using relative simplification. Finally, the operational semantics of function reduction is shown to be congruent with the semantics of $\psi$-terms as approximation structures.

Sort Intersection:
(S.1) $\frac{\psi \& U: s \& U: s^{\prime}}{\psi \& U: s \wedge s^{\prime}}$

## Sort Containment:

(S.2) $\frac{\psi \& U \doteq X \& U: s}{\psi \& U \doteq X}$ if $X: s^{\prime} \in \phi$, and $s^{\prime} \leq s$

Sort Refinement:
(s.3) $\frac{\psi \& U \doteq X \& U: s}{\psi \& U \doteq X \& U: s \wedge s^{\prime}} \quad$ if $X: s^{\prime} \in \phi$, and $s \wedge s^{\prime}<s$

Relative Sort Intersection:
(S.4) $\frac{\psi \& U \doteq X \& U \doteq X^{\prime}}{\psi \& U \doteq X \& U \doteq X^{\prime} \& U: s \wedge s^{\prime}} \begin{array}{ll}i f X: s \in \phi, X^{\prime}: s^{\prime} \in \phi, \\ \text { and } U: s^{\prime \prime} \notin \psi, \text { for any sort } s^{\prime \prime}\end{array}$

Sort Inconsistency:
(s.5) $\frac{\psi \& U: \perp}{\perp}$

Fig. 6. Simplification relatively to $\phi$ : Sorts

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[^0]:    ${ }^{1}$ Several paucrns specifying a same function may possibly have overlapping denotations and therefore the order of the specified patuerns define an implicit priority, as is usual in functional programming using first-order patterns (e.g., [12]).

[^1]:    ${ }^{2}$ Sorts and their relative ordering are specified by the user.

[^2]:    ${ }^{3}$ To illustrate the $\psi$-term ordering, we will give a decreasing matching sequence of $\psi$-terms going from more general to more specific ones.

[^3]:    ${ }^{4}$ Of course, one can use actual values of sort int, real, or string in expressions with their usual operations as in most programming languages. In fact, LIFE provides the additional freedom to write such expressions mixing actual values or their sort approximations int, real, or string. Such expressions are cither solved by local propagation or residuale pending further refinements of the non-value sorts into values.

[^4]:    ${ }^{5}$ Refer to the sorts of Figure 1.

[^5]:    ${ }^{6}$ We assume that poseven $\wedge$ posodd $=1$.

