

$P\rho$ Log: a System for Rule-based Programming

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Outline

Introduction

$P\rho$ Log language

Programming in $P\rho$ Log

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What is $P\rho$ Log

- ▶ A system that extends Prolog with strategic conditional transformation rules.
- ▶ Rules perform nondeterministic transformations of sequences.
- ▶ Strategies provide control on rule applications.
- ▶ $P\rho$ Log system combines the power of logic programming and the flexibility of strategy-based conditional transformation in a single framework.

What is $P\rho$ Log

- ▶ $P\rho$ Log supports programming with four different types of variables: individual, sequence, function and context variables.
- ▶ $P\rho$ Log is expressive enough to specifying and prototyping deductive systems, solvers for various equational theories, tools for XML querying and transformation, etc.
- ▶ $P\rho$ Log code is usually quite short, declaratively clear, and reusable.
- ▶ Implemented in Prolog, available from
<http://www.risc.jku.at/~tkutsia/software.html>

Different Kinds of Variables

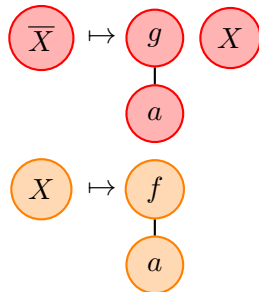
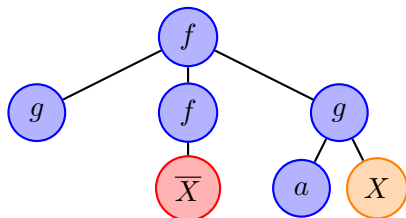
- ▶ Individual variables stand for single terms, while sequence variables stand for finite (possibly empty) sequences of terms.
- ▶ Function variables denote function symbols, while context variables denote contexts that can be seen as unary functions with a single occurrence of the bound variable.
- ▶ Four different types of variables give the user flexibility on selecting subsequences in sequences or subterms/contexts in terms.
- ▶ These variables enhance expressive capabilities of a language, help to write short, neat, understandable code, and hide away many tedious data processing details from the programmer.

Intuition Behind Individual (X) and Sequence Variables (\bar{X})

Example

$$f(g, f(\bar{X}), g(a, X))$$

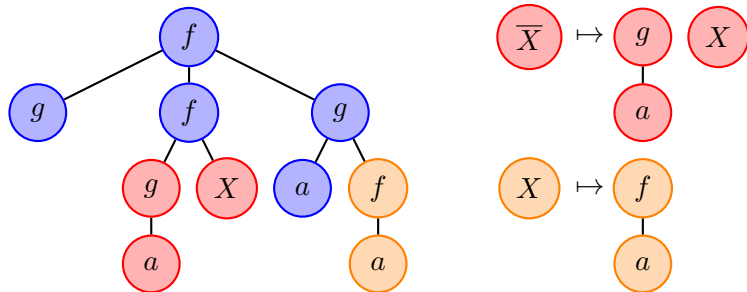
$$\{\bar{X} \mapsto (g(a), X), X \mapsto f(a)\}$$



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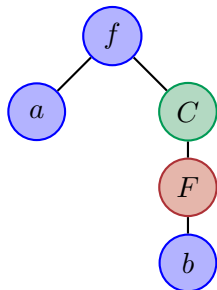
$$f(g, f(g(a), y), g(a, f(a))) \quad \{\bar{X} \mapsto (g(a), X), X \mapsto f(a)\}$$



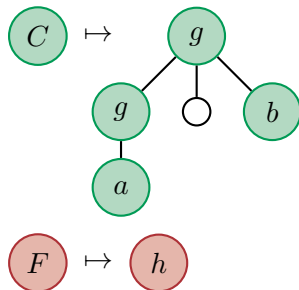
Intuition Behind Function (F) and Context Variables (C)

Example

$$f(a, C(F(b)))$$



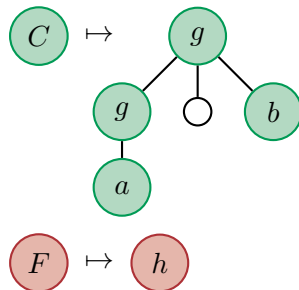
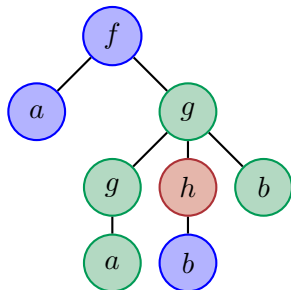
$$\{C \mapsto g(g(a), \circ, b), F \mapsto h\}$$



Intuition Behind Function (F) and Context Variables (C)

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$$f(a, g(g(a), h(b), b)) \quad \{C \mapsto g(g(a), \circ, b), F \mapsto h\}$$



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Terms and Sequences

Terms and sequences are defined as follows:

- ▶ $t ::= X \mid f(s) \mid F(s) \mid C(t)$
- ▶ $s ::= t \mid \bar{X} \mid (s_1, \dots, s_n)$

where

1. X is an individual variable
2. \bar{X} is a sequence variable
3. F is a function variable
4. C is a context variable
5. f is a function symbol

Atoms, Literals

- ▶ ρ -atoms have a form $st :: s_1 \Rightarrow s_2$.
 - ▶ st : strategy (a term).
 - ▶ s_1, s_2 : sequences.
 - ▶ Intuitive meaning: the strategy st transforms the sequence s_1 to the sequence s_2 .
- ▶ Negation of a ρ -atom: $st :: s_1 \not\Rightarrow s_2$.
- ▶ Prolog conventions for naming symbols apply.

Clauses, Queries

- ▶ $P\rho$ Log clauses have a form:

$$st :: s_1 \Rightarrow s_2 :- L_1, \dots, L_n, n \geq 0.$$

- ▶ Each L_i is either a ρ -literal or a Prolog literal.
- ▶ Prolog clauses can be used in $P\rho$ Log programs as well.
- ▶ $P\rho$ Log queries: Conjunction of ρ - or Prolog literals:

$$L_1, \dots, L_n$$

- ▶ Well-modedness.

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 - ▶ $prox(\lambda) :: s_1 \Rightarrow s_2$ succeeds if s_1 matches approximately with s_2 , at least with the degree λ .
 - ▶ etc.

Semantics of P_{ρ} Log

We studied operational and declarative semantics of P_{ρ} Log from constraint logic programming point of view.

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Sorting

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The following program illustrates how bubble sort can be implemented in P ρ Log:

$$\text{swap} :: (\bar{X}, X, Y, \bar{Y}) \Rightarrow (\bar{X}, Y, X, \bar{Y}) :- X > Y.$$
$$\text{sort} :: \bar{X} \Rightarrow \bar{Y} :- \text{nf}(\text{swap}) :: \bar{X} \Rightarrow \bar{Y}.$$

Query:

$$\text{sort} :: (3, 1, 1, 2) \Rightarrow \bar{R}.$$

Outputs: $\bar{R} = (1, 1, 2, 3)$

Example: Merge Proximals

Assume our proximity relation is such that a and b are proximal with the degree 0.6 and b is close to c with the degree 0.8. Then we have:

- ▶ Merge proximals from a sequence:

$$\text{merge_proximals}(\lambda) :: (\bar{X}, X, \bar{Y}, Y, \bar{Z}) \Longrightarrow (\bar{X}, \bar{Y}, Y, \bar{Z}) : -$$
$$\text{prox}(\lambda) :: X \Longrightarrow Y.$$
$$\text{merge_all_proximals}(\lambda) := \text{first_one}(\text{nf}(\text{merge_proximals}(\lambda)))$$

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Applications of $P\rho$ Log

We have applications of $P\rho$ Log in

- ▶ XML processing,
- ▶ Web reasoning,
- ▶ Implementing rewriting strategies,
- ▶ Extraction of frequent patterns from data mining workflows,
- ▶ Modeling of access control policies.

Acknowledgement

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