

# Quantitative Logic Programming Revisited

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# Introduction (I)

**Motivation:** work out an expressive framework for *Constraint Funcional-Logic Programming* with uncertainty.

**First step:** revise an early proposal by *van Emden* (1986): *Quantitative Logic Programming* and improve it in two ways:

- generalizing *van Emden's QLP* to a generic scheme  $QLP(\mathcal{D})$  parametrized by a qualification domain  $\mathcal{D}$ .
- generalizing *van Emden's* results by providing stronger ones, both in declarative semantics and goal solving.

In addition, our  $QLP(\mathcal{D})$  scheme has interesting instances that have been implemented using *CLP*.

## Introduction (and II): *Van Emden's proposal*

- *Van Emden* considered *Quantitative Logic Programs (QLP)* as sets of clauses of the form:

$$A \leftarrow f - B_1, \dots, B_n$$

for which the certainty being propagated to a clause head was  $f \times b$  where:

$f$  - *attenuation factor* with  $f \in (0, 1]$

$b$  - minimum of the certainty factors known for the body atoms

- This approach led to:
  - general results on model theoretic and fixpoint semantics (similar to those for classical LP)
  - procedure for computing the certainty factors in the least Herbrand model of a given program by an alpha-beta heuristic (only for ground atoms with finite search trees)

# Qualification Domains (I): Axioms

- Structure  $\mathcal{D} = \langle D, \sqsubseteq, \perp, \top, \circ \rangle$  such that:
  - $\langle D, \sqsubseteq, \perp, \top \rangle$  is a lattice with extreme points  $\perp$  and  $\top$  w.r.t.  $\sqsubseteq$ .
  - $\circ : D \times D \rightarrow D$ , called *attenuation operation*, verifies:
    - (a)  $\circ$  is associative, commutative and monotonic w.r.t.  $\sqsubseteq$ .
    - (b)  $\forall d \in D : d \circ \top = d$ .
    - (c)  $\forall d \in D : d \circ \perp = \perp$ .
    - (d)  $\forall d, e \in D \setminus \{\perp, \top\} : d \circ e \sqsubseteq e$ .
    - (e)  $\forall d, e_1, e_2 \in D : d \circ (e_1 \sqcap e_2) = d \circ e_1 \sqcap d \circ e_2$ .
- We generalize *van Emden's QLP* to  $QLP(\mathcal{D})$  where:
  - qualification values  $d \in D \setminus \{\perp\}$  instead of  $d \in (0, 1]$ .
  - the *glb* operator  $\sqcap$  instead of  $\min$ .
  - $\circ$  instead of  $\times$  (product).

## Qualification Domains (and II): Instances

- $\mathcal{B} = (\{0, 1\}, \leq, 0, 1, \wedge)$  (Classical Boolean Values).
- $\mathcal{U} = ([0, 1], \leq, 0, 1, \times)$  (*van Emden's* Certainty Degrees).
- $\mathcal{W} = ([0, \infty], \geq, \infty, 0, +)$  (Proof trees' depths).
  
- The cartesian product  $\mathcal{D} = \mathcal{D}_1 \times \mathcal{D}_2$  of two given qualification domains is always another qualification domain.
- Note:  $\mathcal{U} \times \mathcal{W}$  qualifies both certainties and proof trees' depths.

## Syntax (I): Programs

**Program:** set of *qualified definite Horn clauses* of the form

$$A \leftarrow d - B_1, \dots, B_k$$

### Example 1a

```
cruel(X) <-0.90- human(X), eats(X,Y), animal(Y)
```

### Example 1b

```
cruel(X) <-1- human(X), eats(X,Y), animal(Y)
```

### Example 1c

```
cruel(X) <-0.90,1- human(X), eats(X,Y), animal(Y)
```

## Syntax (and II): Interpretations and Models

- $\mathcal{D}$ -annotated atom:  $A \sharp d$ .
- Open  $\mathcal{D}$ -annotated atom:  $A \sharp W$  ( $W$  is intended to take values over  $D \setminus \{\perp\}$ ).

$\mathcal{D}$ -entailment relation: is defined as  $A \sharp d \succ_{\mathcal{D}} A' \sharp d'$  iff there is some substitution  $\theta$  such that  $A' = A\theta$  and  $d' \sqsubseteq d$ .

*Open Herbrand interpretation over  $\mathcal{D}$* : is defined as any set  $\mathcal{I}$  of  $\mathcal{D}$ -annotated atoms which is closed under  $\mathcal{D}$ -entailment. The set of all interpretations over  $\mathcal{D}$  is a complete lattice under  $\sqsubseteq$ .

**Model:** An interpretation  $\mathcal{I}$  is a *model* of a clause  $A \leftarrow d - B_1, \dots, B_k$  in the program  $\mathcal{P}$  iff for any  $\theta$  and any  $d_1, \dots, d_k \in D \setminus \{\perp\}$  such that  $B_i\theta \sharp d_i \in \mathcal{I}$  ( $1 \leq i \leq k$ ), one has  $A\theta \sharp (d \circ \prod\{d_1, \dots, d_k\}) \in \mathcal{I}$ . And we say  $\mathcal{I}$  is a model of  $\mathcal{P}$  if  $\mathcal{I}$  is a model of each clause in  $\mathcal{P}$ .



# Declarative Semantics (I)

**Qualified Horn Logic over  $\mathcal{D}$ :**  $QHL(\mathcal{D})$  is defined as a deductive system consisting just of one inference rule  $QMP(\mathcal{D})$  called *Qualified Modus Ponens* over  $\mathcal{D}$ . If there are some  $(A \leftarrow d - B_1, \dots, B_k) \in \mathcal{P}$ , some substitution  $\theta$  such that  $A' = A\theta$  and  $B'_i = B_i\theta$  for all  $1 \leq i \leq k$  and  $d' \sqsubseteq d \circ \prod\{d_1, \dots, d_k\}$ , the following inference step is allowed:

$$\frac{B'_1 \# d_1 \quad \cdots \quad B'_k \# d_k}{A' \# d'} \quad QMP(\mathcal{D})$$

**Notation**  $\mathcal{P} \vdash_{QHL(\mathcal{D})} A \# d$  or  $\mathcal{P} \vdash_{QHL(\mathcal{D})}^n A \# d$ .

## Declarative Semantics (II)

### Example 2

```
cruel(X) <-0.90- human(X), eats(X,Y), animal(Y)
animal(bird) <-1.0-
human(eve) <-1.0-
human(mother(X)) <-0.90- human(X)
eats(eve,X) <-0.30- animal(X)
eats(mother(X),Y) <-0.70- eats(X,Y)
```

# Declarative Semantics (II)

## Example 2

```
cruel(X) <-0.90- human(X), eats(X,Y), animal(Y)
animal(bird) <-1.0-
human(eve) <-1.0-
human(mother(X)) <-0.90- human(X)
eats(eve,X) <-0.30- animal(X)
eats(mother(X),Y) <-0.70- eats(X,Y)
```

<u>human(eve)#1.0</u>	<u>animal(bird)#1.0</u>	
	<u>eats(eve,bird)#0.30</u>	
<u>human(mother(eve))#0.90</u>	<u>eats(mother(eve),bird)#0.21</u>	<u>animal(bird)#1.0</u>
	<u>cruel(mother(eve))#0.15</u>	

# Declarative Semantics (II)

## Example 2

```
cruel(X) <-0.90- human(X), eats(X,Y), animal(Y)
animal(bird) <-1.0-
human(eve) <-1.0-
human(mother(X)) <-0.90- human(X)
eats(eve,X) <-0.30- animal(X)
eats(mother(X),Y) <-0.70- eats(X,Y)
```

<u>human(eve)#1.0</u>	<u>animal(bird)#1.0</u> <u>eats(eve,bird)#0.30</u>	<u>animal(bird)#1.0</u>
human(mother(eve))#0.90	eats(mother(eve),bird)#0.21	*
<hr style="border: 0.5px solid black;"/>		
cruel(mother(eve))#0.15		

(\*)  $0.15 \leq 0.90 * \min \{0.90, 0.21, 1.0\} = 0.189$  .

# Declarative Semantics (and III)

Characterizations of the least Herbrand model:

- 1 The least fixpoint  $\mu(\mathbb{T}_{\mathcal{P}})$  is the least Herbrand model of  $\mathcal{P}$ , noted as  $\mathcal{M}_{\mathcal{P}}$ .
- 2  $\mathcal{M}_{\mathcal{P}} = \{A \# d \mid \mathcal{P} \vdash_{\text{QHL}(\mathcal{D})} A \# d\}$ .

# Goals

- Initial goals:

$$\dots, A_i \# W_i, \dots \quad \sqsubseteq \in \sqsubseteq \quad \dots, W_i \sqsupseteq \beta_i, \dots$$

- General goals:

$$\dots, A_i \# W_i, \dots \quad \sqsubseteq \sigma \sqsubseteq \quad \dots, W' = d \circ \bigsqcap \{W'_1, \dots, W'_k\},$$
$$\dots, \alpha \circ W_i \sqsupseteq \beta_i, \dots$$

- Solved goals:

$$\sqsubseteq \sigma \sqsubseteq \quad \dots, W' = d \circ \bigsqcap \{W'_1, \dots, W'_k\}, \dots$$

# SLD( $\mathcal{D}$ ) Resolution (I)

- Resolution step:

$$\dots, A \# W, \dots \Vdash \sigma \Vdash \dots, \alpha \circ W \sqsupseteq \beta, \dots \Vdash_{C_1, \sigma_1}$$

$$\begin{aligned} &(\dots, B_1 \# W_1, \dots, B_k \# W_k, \dots) \sigma_1 \Vdash \sigma \sigma_1 \Vdash \dots, \\ & d \circ \alpha \circ W_1 \sqsupseteq \beta, \dots, d \circ \alpha \circ W_k \sqsupseteq \beta, \dots, \\ & W = d \circ \bigsqcap \{W_1, \dots, W_k\}, \dots \end{aligned}$$

where  $C_1 \equiv (H \leftarrow d - B_1, \dots, B_k) \in \mathcal{P}$  such that  $d \circ \alpha \sqsupseteq \beta$  and  $\sigma_1$  the m.g.u. between  $A$  and  $H$ .

- Resolution computations:

$$G_0 \Vdash_{\sigma_1} G_1 \Vdash_{\sigma_2} \dots \Vdash_{\sigma_n} G_n \equiv \sigma_1 \sigma_2 \dots \sigma_n \Vdash \Delta_n$$

with  $G_0$  initial and  $G_n$  solved, and the computed answer  $(\sigma, \mu)$  where  $\sigma = \sigma_1 \sigma_2 \dots \sigma_n$  and  $\mu(W)$  computed according to  $\Delta_n$ .

# SLD( $\mathcal{D}$ ) Resolution (and II)

## Example 3

`eats(X,Y)#W | {} | W >= 0.20`

$\Vdash_{\text{eats}.2, \{X \mapsto \text{mother}(X')\}}$



# SLD( $\mathcal{D}$ ) Resolution (and II)

## Example 3

$\text{eats}(X,Y)\#W \mid \{\} \mid W \geq 0.20$

$\text{eats}(X',Y)\#W1 \mid \{X \mapsto \text{mother}(X')\} \mid$

$W = 0.70 * \min \{W1\},$

$0.70 * W1 \geq 0.20$

$\Vdash_{\text{eats}.2, \{X \mapsto \text{mother}(X')\}}$

$\Vdash_{\text{eats}.1, \{X' \mapsto \text{eve}\}}$

# SLD( $\mathcal{D}$ ) Resolution (and II)

## Example 3

$\text{eats}(X,Y)\#W \mid \{\} \mid W \geq 0.20$	$\Vdash_{\text{eats}.2, \{X \mapsto \text{mother}(X')\}}$
$\text{eats}(X',Y)\#W1 \mid \{X \mapsto \text{mother}(X')\} \mid$ $W = 0.70 * \min \{W1\},$ $0.70 * W1 \geq 0.20$	$\Vdash_{\text{eats}.1, \{X' \mapsto \text{eve}\}}$
$\text{animal}(Y)\#W2 \mid \{X \mapsto \text{mother}(\text{eve})\} \mid$ $W = 0.70 * \min \{W1\},$ $W1 = 0.30 * \min \{W2\},$ $0.30 * 0.70 * W2 \geq 0.20$	$\Vdash_{\text{animal}, \{Y \mapsto \text{bird}\}}$

# SLD( $\mathcal{D}$ ) Resolution (and II)

## Example 3

eats(X,Y)#W   {}   W >= 0.20	$\Vdash_{\text{eats}.2, \{X \mapsto \text{mother}(X')\}}$
eats(X',Y)#W1   {X $\mapsto$ mother(X')}   W = 0.70 * min {W1}, 0.70 * W1 >= 0.20	$\Vdash_{\text{eats}.1, \{X' \mapsto \text{eve}\}}$
animal(Y)#W2   {X $\mapsto$ mother(eve)}   W = 0.70 * min {W1}, W1 = 0.30 * min {W2}, 0.30 * 0.70 * W2 >= 0.20	$\Vdash_{\text{animal}, \{Y \mapsto \text{bird}\}}$
{X $\mapsto$ mother(eve), Y $\mapsto$ bird}   W = 0.70 * min {W1}, W1 = 0.30 * min {W2}, W2 = 1.0	

# SLD( $\mathcal{D}$ ) Resolution (and II)

## Example 3

eats(X,Y)#W   {}   W >= 0.20	$\Vdash_{\text{eats.2},\{X \mapsto \text{mother}(X')\}}$
eats(X',Y)#W1   {X $\mapsto$ mother(X')}	
W = 0.70 * min {W1},	
0.70 * W1 >= 0.20	$\Vdash_{\text{eats.1},\{X' \mapsto \text{eve}\}}$
animal(Y)#W2   {X $\mapsto$ mother(eve)}	
W = 0.70 * min {W1},	
W1 = 0.30 * min {W2},	
0.30 * 0.70 * W2 >= 0.20	$\Vdash_{\text{animal},\{Y \mapsto \text{bird}\}}$
{X $\mapsto$ mother(eve), Y $\mapsto$ bird}	
W = 0.70 * min {W1},	
W1 = 0.30 * min {W2},	
W2 = 1.0	

Computed answer:  $\{X \mapsto \text{mother}(\text{eve}), Y \mapsto \text{bird}\} \sqcap \{W \mapsto 0.21\}$

# Properties of SLD( $\mathcal{D}$ ) Resolution

- **Definition of Solution:** A pair of substitutions  $(\theta, \rho)$  is called solution of a goal  $G \equiv \bar{A} \parallel \sigma \parallel \Delta$  iff:
  - (i)  $\theta = \sigma\theta$ .
  - (ii)  $\rho \in \text{Sol}(\Delta)$ .
  - (iii)  $\mathcal{P} \vdash_{\text{QHL}(\mathcal{D})} A\theta \# W\rho$  for every annotated atom in  $\bar{A}$ .

**Soundness:** Assume  $G_0 \Vdash^* G$  and  $G = \sigma \parallel \Delta$  solved. Let  $(\sigma, \mu)$  be the computed answer to  $G$ . Then  $(\sigma, \mu)$  is a solution of  $G_0$ .

**Strong Completeness:** Assume a given solution  $(\theta, \rho)$  for  $G_0$  and any fixed strategy for choosing the selected atom at each resolution step. Then there is some computed answer  $(\sigma, \mu)$  for  $G_0$  which is a more general solution than  $(\theta, \rho)$ .

# Implementation: translating to $\mathcal{TOY}$

**Requirement**  $CLP$  or  $CFLP$  system with support for  $\mathcal{C}_D$  constraints.

$$C \equiv a(\bar{t}) \leftarrow d - b_1(\bar{s}_1), \dots, b_k(\bar{s}_k)$$

$$C^t \equiv a(\bar{t}, Alpha, W, Beta) \leftarrow \begin{array}{l} d \circ Alpha \sqsupseteq Beta, \\ b_1(\bar{s}_1, d \circ Alpha, W_1, Beta), \\ \vdots \\ b_k(\bar{s}_k, d \circ Alpha, W_k, Beta), \\ W = d \circ \prod \{W_1, \dots, W_k\} \end{array}$$

$$G \equiv a_1(\bar{t}_1) \# W_1, \dots, a_m(\bar{t}_m) \# W_m \sqcap W_1 \sqsupseteq \beta_1, \dots, W_m \sqsupseteq \beta_m$$

$$G^t \equiv a_1(\bar{t}_1, \top, W_1, \beta_1), \dots, a_m(\bar{t}_m, \top, W_m, \beta_m)$$

# Conclusions

- Generalization of the QLP proposal by *van Emden* to a generic scheme  $QLP(\mathcal{D})$  covering uncertainty and more.
- Sound and complete goal solving procedure improving *van Emden's QLP*.
- Implementation of some  $QLP(\mathcal{D})$  instances by means of a translation to  $CLP(\mathcal{C}_D)$  realized in *TOY*.

## Future work:

- Simple treatment of negation.
- Threshold constraints in clause bodies.
- Similarity-based reasoning.
- Extension with lazy functions and constraints.