



ANR Chair of Excellence

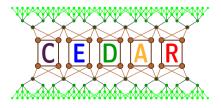
**Université Claude Bernard Lyon 1** 







Constraint Event-Driven Automated Reasoning Project



#### **Outline**

- Constraint Logic Programming
- What is unification?
- Semantic Web objects
- Graphs as constraints
- $ightharpoonup \mathcal{OWL}$  and  $\mathcal{DL}$ -based reasoning
- Constraint-based Semantic Web reasoning
- Recapitulation

#### **Outline**

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## **Constraint Logic Programming**

In Prolog seen as a CLP language, a clause such as:

```
append([],L,L).

append([H|T],L,[H|R]) :- append(T,L,R).
```

#### is construed as:

## **Constraint Logic Programming Scheme**

The  $\mathcal{CLP}$  scheme requires a set  $\mathcal{R}$  of relational symbols (or, predicate symbols) and a constraint language  $\mathcal{L}$ .

The constraint language  $\mathcal{L}$  needs very little —(not even syntax!):

- ightharpoonup a set  $\mathcal{V}$  of *variables* (denoted as capitalized  $X,Y,\ldots$ );
- ▶ a set  $\Phi$  of *formulae* (denoted  $\phi, \phi', ...$ ) called *constraints*;
- ▶ a function VAR:  $\Phi \mapsto \mathcal{V}$ , giving for every constraint  $\phi$  the set VAR( $\phi$ ) of *variables constrained by*  $\phi$ ;
- ▶ a family of *interpretations*  $\mathcal{A}$  over some domain  $D^{\mathcal{A}}$ ;
- ▶ a set VAL(A) of *valuations*—total functions  $\alpha : V \mapsto D^A$ .

## **Constraint Logic Programming Language**

Given a set of relational symbols  $\mathcal{R}$   $(r, r_1, ...)$ , a constraint language  $\mathcal{L}$  is extended into a language  $\mathcal{R}(\mathcal{L})$  of *constrained relational clauses* with:

- ▶ the set  $\mathcal{R}(\Phi)$  of formulae defined to include:
  - all formulae  $\phi$  in  $\Phi$ , *i.e.*, all  $\mathcal{L}$ -constraints;
  - all relational atoms  $r(X_1, \ldots, X_n)$ , where  $X_1, \ldots, X_n \in \mathcal{V}$  are mutually distinct;
  - and closed under & (conjunction) and  $\rightarrow$  (implication);
- ightharpoonup extending an interpretation  $\mathcal{A}$  of  $\mathcal{L}$  by adding relations:  $r^{\mathcal{A}} \subset D^{\mathcal{A}} \times \ldots \times D^{\mathcal{A}}$  for each  $r \in \mathcal{R}$ .

## **Constraint Logic Programming Clause**

We define a CLP constrained *definite clause* in R(L) as:

$$r(\vec{X}) \leftarrow r_1(\vec{X}_1) \& \ldots \& r_m(\vec{X}_m) \ \ \phi,$$

where  $(0 \le m)$  and:

- $ightharpoonup r(\vec{X}), r_1(\vec{X}_1), \ldots, r_m(\vec{X}_m)$  are relational atoms in  $\mathcal{R}(\mathcal{L})$ ; and,
- $ightharpoonup \phi$  is a constraint formula in  $\mathcal{L}$ .

A constrained *resolvent* is a formula  $\varrho \mid \phi$ , where  $\varrho$  is a (possibly empty) conjunction of relational atoms  $r(X_1,\ldots,X_n)$ —its *relational part*—and  $\varphi$  is a (possibly empty) conjunction of  $\mathcal{L}$ -constraints—its *constraint part*.

## **Constraint Logic Programming Resolution**

Constrained *resolution* is a reduction rule on resolvents that gives a sound and complete interpreter for *programs* consisting of a set  $\mathcal{C}$  of constrained definite  $\mathcal{R}(\mathcal{L})$ -clauses.

The reduction of a constrained *resolvent* of the form:

$$B_1 \& \ldots \& r(X_1,\ldots,X_n) \& \ldots B_k \mid \phi$$

by the (renamed) program clause:

$$r(X_1,\ldots,X_n) \leftarrow A_1 \& \ldots \& A_m \parallel \phi'$$

is the new constrained resolvent of the form:

$$B_1 \& \ldots \& A_1 \& \ldots \& A_m \& \ldots B_k \mid \phi \& \phi'.$$

# Some important points:

- ► But... wait a minute: "Constraints are logical formulae—so why not use only logic?"
  - Indeed, constraints are logical formulae—and that is *good!*But such formulae as factors in a conjunction *commute* with other factors, thus freeing operational scheduling of resolvents.
- A constraint is a formula solvable by a specific solving algorithm rather than general-purpose logic-programming machinery.
- ▶ Better: constraint solving remembers proven facts (proof memoizing).

Such are key points exploited in CLP!

## **Constraint Solving—Constraint Normalization**

Constraint solving is conveniently specified using *constraint normalization rules*, which are semantics-preserving syntax-driven rewrite (meta-)rules.

Plotkin's SOS notation:

A normalization rule is said to be *correct* iff the prior form's denotation is equal to the posterior form's whenever the side condition holds.

## **Constraint Normalization—Declarative Coroutining**

Normalizing a constraint yields a **normal form**: a constraint formula that can't be transformed by any normalization rule.

Such may be either the inconsistent constraint  $\perp$ , or:

- ➤ a solved form—a normal form that can be immediately deemed consistent; or,
- ▶ a residuated form—a normal form but not a solved form.

A residuated constraint is a *suspended* computation; shared variables are inter-process communication channels: binding in one normalization process may trigger resumption of another residuated normalization process.

Constraint residuation enables automatic coroutining!

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#### What is unification?—First-order terms

The set  $\mathcal{T}_{\Sigma,\mathcal{V}}$  of *first-order terms* is defined given:

- ▶ V a countable set of variables;
- $ightharpoonup \Sigma_n$  sets of *constructors* of arity  $n \ (n \ge 0)$ ;
- $ightharpoonup \Sigma = \bigcup_{n>0} \Sigma_n$  the constructor *signature*.

Then, a first-order term (FOT) is either:

- $\triangleright$  a variable in  $\mathcal{V}$ ; or,
- ightharpoonup an element of  $\Sigma_0$ ; or,
- ▶ an expression of the form  $f(t_1, ..., t_n)$ , where n > 0,  $f \in \Sigma_n$ , and  $t_i$  is a FOT, for all  $i \ge 1$ .

Examples of FOTs: X a f(g(X, a), Y, h(X)) (variables are capitalized as in Prolog).

#### What is unification?—Substitutions & instances

A variable substitution is a map  $\sigma: \mathcal{V} \to \mathcal{T}_{\Sigma,\mathcal{V}}$  such that the set  $\{X \in \mathcal{V} \mid \sigma(X) \neq X\}$  is finite.

Given a substitution  $\sigma$  and a FOT t, the  $\sigma$ -instance of t is the FOT:

$$t\sigma = \begin{cases} \sigma(X) & \text{if } t = X \in \mathcal{V}; \\ a & \text{if } t = a \in \Sigma_0; \\ f(t_1\sigma, \dots, t_n\sigma) & \text{if } t = f(t_1, \dots, t_n). \end{cases}$$

**Unification** is the process of solving an equation of the form:

$$t \doteq t'$$

## What is unification?—FOT equation solving

A **solution**, if one exists, is any substitution  $\sigma$  such that:

$$t\sigma = t'\sigma$$

If solutions exist, there is always a **minimal** solution (<u>the</u> most general unifier): mgu(t, t').

where: " $\sigma_1$  is more general than  $\sigma_2$ " iff  $\exists \sigma$  s.t.  $\sigma_2 = \sigma_1 \sigma$ 

# **Equation and solution example:**

$$f(g(X,b),X,g(h(X),Y)) \doteq f(g(U,U),b,g(V,a))$$
 
$$X \doteq b,Y \doteq a,U \doteq b,V \doteq h(b)$$

## What is unification?—Algorithms

FOT unification algorithms have been (re-)invented:

- ▶ J. Herbrand (PhD thesis—page 148, 1930)
- ► J.A. Robinson (JACM 1965)
- ► A. Martelli & U. Montanari (ACM TOPLAS 1982)

But, rather than a monolithic algorithm, FOT unification is simply expressible as a set of syntax-driven **commutative** and terminating constraint normalization rules!

#### What is unification?—Constraint normalization rules

### (1) Substitute

$$\phi \& X \doteq t$$

$$\phi[X/t] \& X \doteq t$$

if X occurs in  $\phi$ 

## (2) Decompose

$$\frac{\phi \& f(s_1,\ldots,s_n) \doteq f(t_1,\ldots,t_n)}{\phi \& s_1 \doteq t_1 \& \ldots \& s_n \doteq t_n} \quad \text{if} \quad f \in \Sigma_n, \ (n \geq 0)$$

(3) Fail

#### What is unification?—Constraint normalization rules

$$\begin{array}{c|cccc} \phi & \& & t \doteq X \\ \hline \phi & \& & X \doteq t \end{array} \quad \text{if} \quad X \in \mathcal{V} \\ \text{and} \quad t \not \in \mathcal{V} \end{array}$$

# (5) Erase

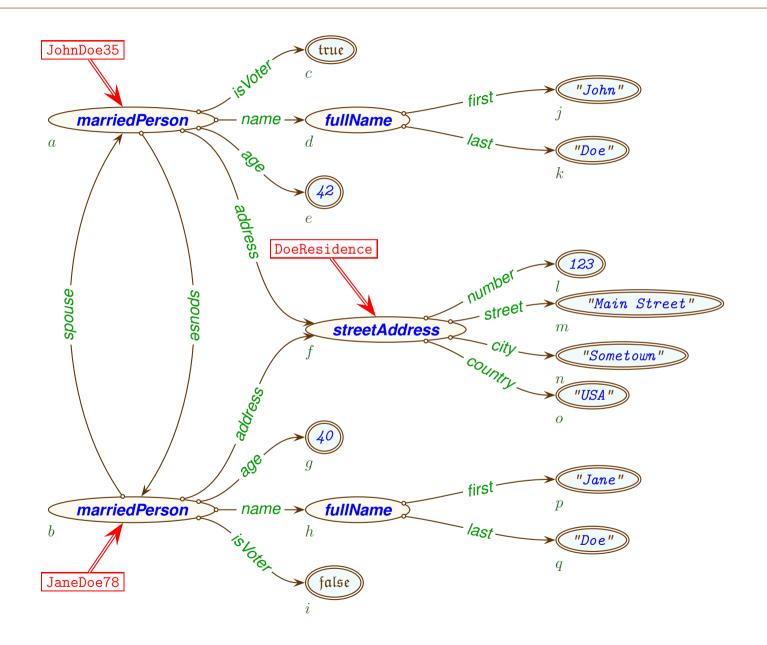
$$\frac{\phi \& t \doteq t}{\phi} \quad \text{if} \ t \in \Sigma_0 \cup \mathcal{V}$$

# (6) Cycle

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## **Semantic Web objects—**Objects are labelled graphs!

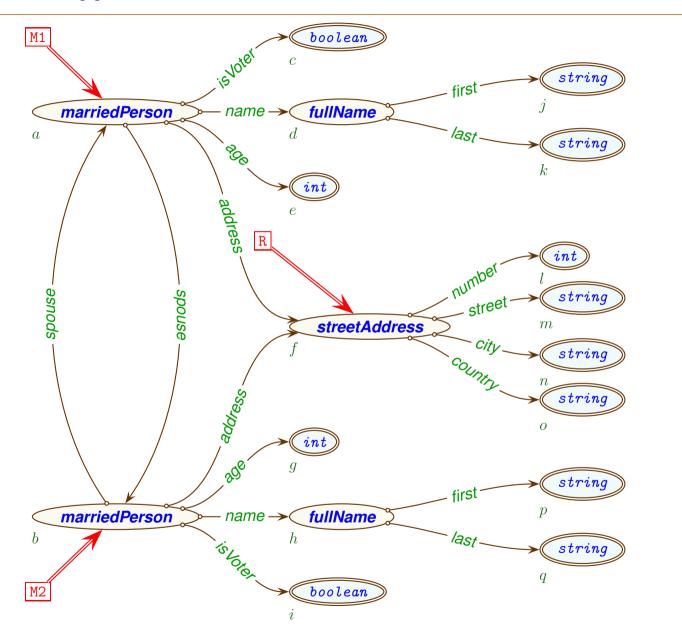


# **Semantic Web objects—**Objects are labelled graphs!

## Semantic Web objects—Objects are labelled graphs!

```
JaneDoe78: marriedPerson ( name => fullName
                                     ( first => "Jane"
                                     , last => "Doe" )
                         , age => 40
                         , address => DoeResidence
                         , spouse => JohnDoe35
                         , isVoter => false
DoeResidence : streetAddress ( number => 123
                              , street => "Main Street"
                              , city => "Sometown"
                              , country => "USA"
```

# **Semantic Web types—**Types are labelled graphs!



## **Semantic Web types—***Types are labelled graphs!*

## Semantic Web formalisms—Types are labelled graphs!

```
M2 : marriedPerson ( name => fullName
                                     ( first => string
                                     , last => string )
                    , age => int
                    , address \Rightarrow R
                    , spouse \Rightarrow M1
                    , isVoter => boolean
R : streetAddress ( number => int
                     , street => string
                     , city => string
                     , country => string
```

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# Original motivation: Formalize this?—ca. 1982

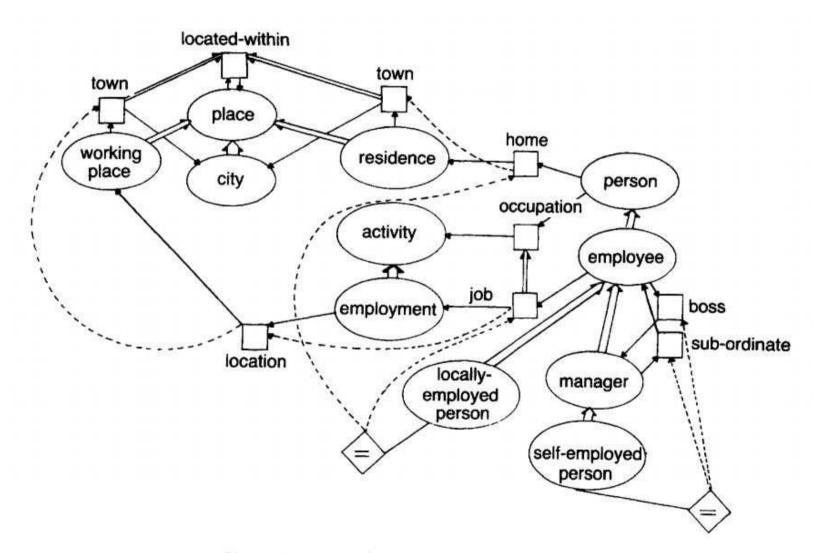


Fig. 1. Example of a KL-ONE semantic network.

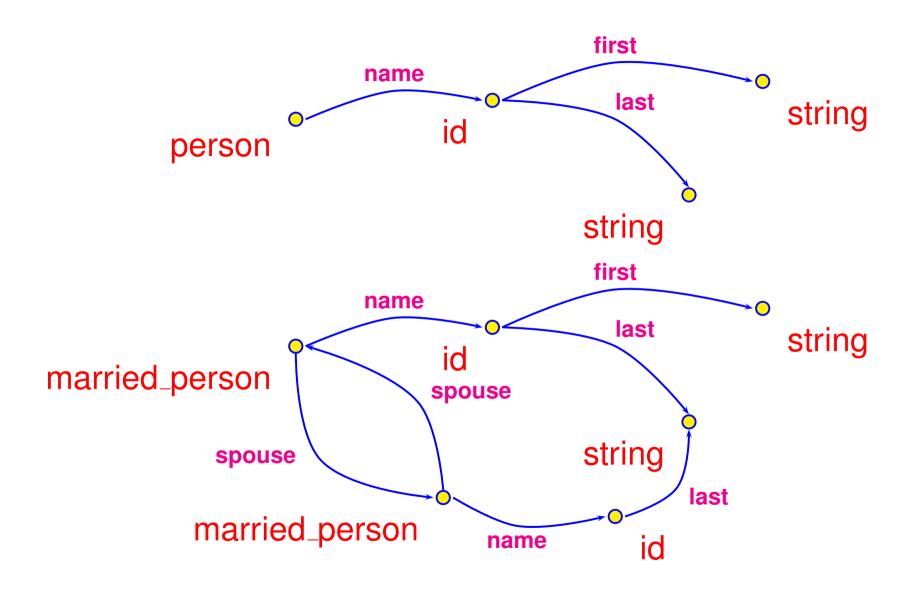
- ➤ What: a formalism for representing objects that is: intuitive (objects as labelled graphs), expressive ("real-life" data models), formal (logical semantics), operational (executable), & efficient (constraint-solving)
- ► Why? viz., ubiquitous use of labelled graphs to structure information naturally as in:
  - object-orientation, knowledge representation,
  - databases, semi-structured data,
  - natural language processing, graphical interfaces,
  - concurrency and communication,
  - XML, RDF, the "Semantic Web," etc., ...

## **Graphs as constraints—***History*

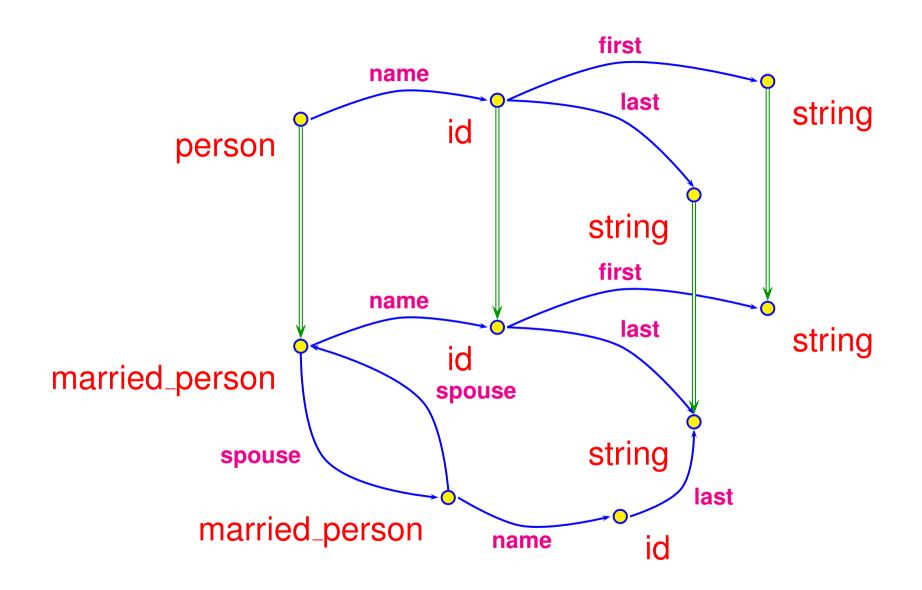
# Viewing graphs as *constraints* stems from the work of:

- ► Hassan Aït-Kaci (since 1983)
- ► Gert Smolka (since 1986)
- ► Andreas Podelski (since 1989)
- ► Franz Baader, Rolf Backhofen, Jochen Dörre, Martin Emele, Bernhard Nebel, Joachim Niehren, Ralf Treinen, Manfred Schmidt-Schauß, Remi Zajac, ...

# Graphs as constraints—Inheritance as graph endomorphism



# Graphs as constraints—Inheritance as graph endomorphism



## Graphs as constraints—OSF term syntax

Let  $\mathcal{V}$  be a countable set of variables, and  $\mathcal{S}$  a lattice of sorts.

An OSF term is an expression of the form:

$$X: s(\ell_1 \Rightarrow t_1, \dots, \ell_n \Rightarrow t_n)$$

#### where:

- $X \in \mathcal{V}$  is the root variable
- $\gt s \in \mathcal{S}$  is the root sort
- $\blacktriangleright \{\ell_1, \ldots, \ell_n\} \subseteq \mathcal{F}$  are features
- $ightharpoonup t_1, \ldots, t_n$  are  $\mathcal{OSF}$  terms

## **Graphs as constraints**—OSF *term syntax example*

```
 \begin{split} X: person(name \Rightarrow N: \top (first \Rightarrow F: string), \\ name \Rightarrow M: id(last \Rightarrow S: string), \\ spouse \Rightarrow P: person(name \Rightarrow I: id(last \Rightarrow S: \top), \\ spouse \Rightarrow X: \top). \end{split}
```

# Lighter notation (showing only shared variables):

```
\begin{array}{l} X: person(name \Rightarrow \top(first \Rightarrow string), \\ name \Rightarrow id(last \Rightarrow S: string), \\ spouse \Rightarrow person(name \Rightarrow id(last \Rightarrow S), \\ spouse \Rightarrow X)). \end{array}
```

### **Graphs as constraints**— $\mathcal{OSF}$ clause syntax

An OSF constraint is one of:

$$X : s$$

$$X \cdot \ell \doteq X'$$

$$X = X'$$

where X(X') is a variable (*i.e.*, a node), s is a sort (*i.e.*, a node's type), and  $\ell$  is a feature (*i.e.*, an arc).

An OSF clause is a conjunction of OSF constraints—*i.e.*, a set of OSF constraints



### Graphs as constraints—From OSF terms to OSF clauses

An  $\mathcal{OSF}$  term  $t = X : s(\ell_1 \Rightarrow t_1, \dots, \ell_n \Rightarrow t_n)$  is dissolved into an  $\mathcal{OSF}$  clause  $\phi(t)$  as follows:

$$\varphi(t) \stackrel{\text{\tiny def}}{=\!\!\!=\!\!\!=} X: s \quad \& \quad X.\ell_1 \doteq X_1 \quad \& \quad \dots \quad \& \quad X.\ell_n \doteq X_n$$

$$\& \quad \varphi(t_1) \qquad \& \quad \dots \quad \& \quad \varphi(t_n)$$

where  $X_1, \ldots, X_n$  are the root variables of  $t_1, \ldots, t_n$ .

### Graphs as constraints—Example of OSF term dissolution

```
t = X : person(name \Rightarrow N : \top(first \Rightarrow F : string),
                  name \Rightarrow M : id(last \Rightarrow S : string),
                  spouse \Rightarrow P: person(name \Rightarrow I: id(last \Rightarrow S: \top),
                                          spouse \Rightarrow X : \top)
\varphi(t) = X : person \& X. name \doteq N \& N: \top
                      & X. name \doteq M & M: id
                      & X. spouse = P & P: person
                      & N. first \doteq F \& F: string
                      & M. last \doteq S & S: string
                      & P.name \doteq I & I:id
                      & I.last \doteq S & S: \top
                      & P.spouse = X & X: \top
```

#### Graphs as constraints—Basic OSF term normalization

$$\phi \& X : s \& X : s' \qquad \phi \& X \stackrel{.}{=} X'$$

$$\phi \& X : s \wedge s'$$

#### (1) Sort Intersection (3) Variable Elimination

#### (2) Inconsistent Sort

$$\phi \& X : \bot$$

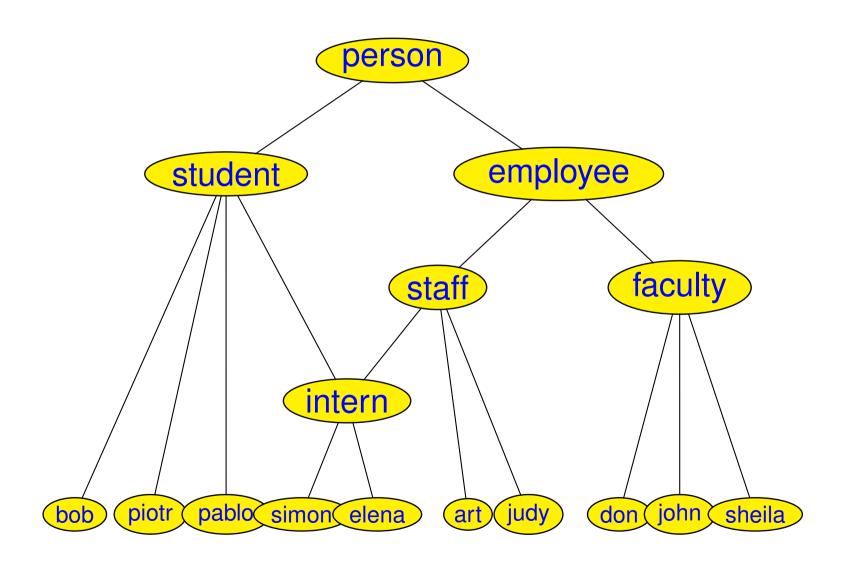
$$X:\bot$$

#### (4) Feature Functionality

$$\phi \& X.\ell \doteq X' \& X.\ell \doteq X''$$

$$\phi \& X.\ell \doteq X' \& X' \doteq X''$$

#### **Graphs as constraints—** $\mathcal{OSF}$ unification as $\mathcal{OSF}$ constraint normalization



#### Graphs as constraints—OSF unification as OSF constraint normalization

```
X : student
      (roommate => person(rep => E : employee),
       advisor => don(secretary => E))
&
  Y : employee
      (advisor => don(assistant => A),
       roommate => S : student(rep => S),
       helper => simon(spouse => A))
&
```

X = Y

#### Graphs as constraints—OSF unification as OSF constraint normalization

```
X : intern
    (roommate => S : intern(rep => S),
     advisor => don(assistant => A,
                    secretary => S),
     helper => simon(spouse => A))
X = Y
E = S
```

&

&

#### **Graphs as constraints—***Extended* OSF *terms*

## Basic OSF terms may be extended to express:

- Non-lattice sort signatures
- Disjunction
- Negation
- Partial features
- Extensional sorts (i.e., denoting elements)
- ► Relational features (a.k.a., "roles")
- Aggregates (à la monoid comprehensions)
- Regular-expression feature paths
- ▶ Sort definitions (a.k.a., "OSF theories"—"ontologies")

#### **Order-sorted featured graph constraints—**(Summary)

We have overviewed a formalism of objects where:

- "real-life" objects are viewed as logical constraints
- objects may be approximated as set-denoting constructs
- object normalization rules provide an efficient operational semantics
- consistency extends unification (and thus matching)
- ▶ this enables rule-based computation (whether rewrite or logical rules) over general graph-based objects
- this yield a powerful means for effectively using ontologies

#### **Reasoning and the Semantic Web**

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#### Semantic Web formalisms— $\mathcal{OWL}$ speaks

What language(s) do OWL's speak?—a confusing growing crowd of strange-sounding languages and logics:

- OWL, OWL Lite, OWL DL, OWL Full
- $\bullet \mathcal{DL}, \mathcal{DLR}, \dots$
- AL, ALC, ALCN, ALCNR, ...
- SHIF, SHIN, CIQ, SHIQ, SHOQ(D), SHOIQ, SRIQ, SROIQ, . . .

## Depending on whether the system allows:

- concepts, roles (inversion, composition, inclusion, ...)
- individuals, datatypes, cardinality constraints
- various combination thereof

For better or worse, the W3C has married its efforts to  $\mathcal{DL}$ -based reasoning systems:

- All the proposed  $\mathcal{DL}$  Knowledge Base formalisms in the  $\mathcal{OWL}$  family use tableaux-based methods for reasoning
- Tableaux methods work by building models explicitly via formula expansion rules
- ▶ This limits  $\mathcal{DL}$  reasoning to finite (*i.e.*, decidable) models
- Worse, tableaux methods only work for small ontologies: they fail to scale up to large ontologies

#### Semantic Web formalisms—DL dialects

## Tableaux style DL reasoning (ALCNR)

#### **CONJUNCTIVE CONCEPT:**

$$\frac{S}{S \cup \{x : C_1, x : C_2\}}$$

#### **EXISTENTIAL ROLE:**

$$\begin{array}{c} \text{if} \quad x: (C_1\sqcap C_2) \in S \\ \text{and} \quad \{x: C_1, x: C_2\} \not\subseteq S \end{array} \end{array} \right] \qquad \begin{array}{c} S \\ \hline S \cup \{x: C_1, x: C_2\} \end{array} \qquad \begin{bmatrix} \text{if} \quad x: (\exists R.C) \in S \text{ s.t. } R \stackrel{\texttt{def}}{=} (\sqcap_{i=1}^m R_i) \\ \text{and} \quad z: C \in S \Rightarrow z \not\in R_S[x] \end{bmatrix} \qquad \begin{array}{c} S \\ \hline S \cup \{xR_iy\}_{i=1}^m \cup \{y: C\} \end{array}$$

$$\frac{S}{S \cup \{xR_iy\}_{i=1}^m \cup \{y:C\}}$$

#### **DISJUNCTIVE CONCEPT:**

$$\left[\begin{array}{cc} \text{if} & x:(C_1\sqcup C_2)\,\in\,S\\ \text{and} & x:C_i\,\not\in\,S\ (i=1,2) \end{array}\right] \qquad \frac{S}{S\cup\left\{x:C_i\right\}}$$

$$\frac{S}{S \cup \{x : C_i\}}$$

#### MIN CARDINALITY:

$$\left[ \begin{array}{ccc} \text{if} & x: (\geq n.R) \in S \text{ s.t. } R \stackrel{\texttt{DEF}}{=\!=\!=} \left( \bigcap_{i=1}^m R_i \right) \\ \text{and} & |R_S[x]| \neq n \\ \text{and} & y_i \text{ is new } (0 \leq i \leq n) \end{array} \right] \qquad \frac{S}{S \cup \left\{ x R_i y_j \right\}_{i,j=1,1}^{m,n} }$$

$$S = S$$

$$S \cup \{xR_iy_j\}_{i,j=1,1}^{m,n}$$

$$\cup \{y_i \neq y_j\}_{1 \leq i < j \leq n}$$

#### **UNIVERSAL ROLE:**

$$\begin{bmatrix} & \text{if} \quad x: (\forall R.C) \in S \\ & \text{and} \quad y \in R_S[x] \\ & \text{and} \quad y: C \not\in S \end{bmatrix}$$

$$\frac{S}{S \cup \{y : C\}}$$

#### MAX CARDINALITY:

$$\left[\begin{array}{cccc} \textbf{if} & x: (\leq n.R) \in S \\ \textbf{and} & |R_S[x]| > n \quad \textbf{and} \quad y, z \in R_S[x] \\ \textbf{and} & y \neq z \not \in S \end{array}\right] \qquad \frac{S}{S \cup S[y/z]}$$

Understanding  $\mathcal{OWL}$  amounts to reasoning with knowledge expressed as  $\mathcal{OWL}$  sentences. Its  $\mathcal{DL}$  semantics relies on explicitly building models using induction.

## ergo:

Inductive techniques are *eager* and (thus) *wasteful* 

Reasoning with knowledge expressed as constrained ( $\mathcal{OSF}$ ) graphs relies on implicitly pruning inconsistent elements using coinduction.

## ergo:

Coinductive techniques are *lazy* and (thus) *thrifty* 

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#### LIFE—Rules + constraints for Semantic Web reasoning

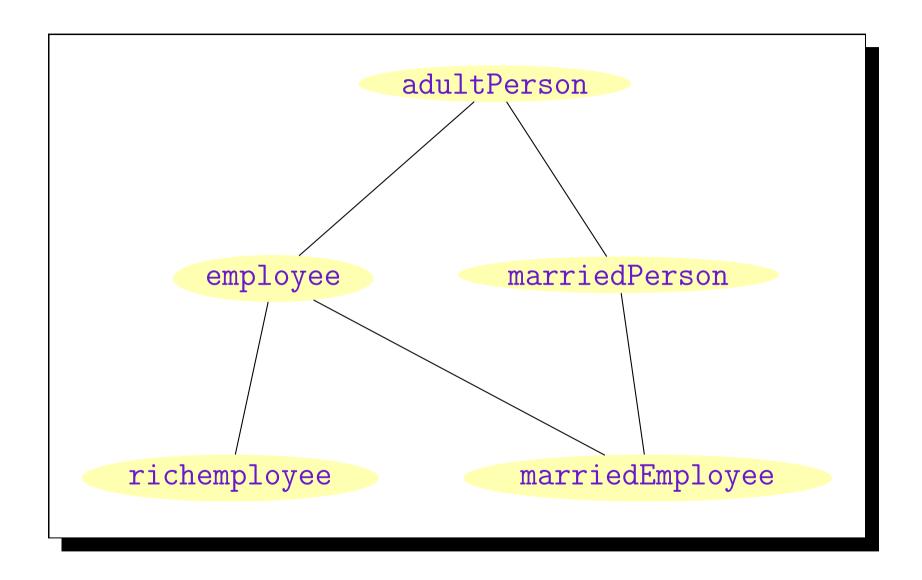
 $\mathcal{LIFE}$ — $\mathcal{L}$ ogic,  $\mathcal{I}$ nheritance,  $\mathcal{F}$ unctions, and  $\mathcal{E}$ quations

 $\mathcal{CLP}(\chi)$ — $\mathcal{C}$ onstraint,  $\mathcal{L}$ ogic,  $\mathcal{P}$ rogramming, parameterized over is a constraint system  $\chi$ 

 $\mathcal{LIFE}$  is a  $\mathcal{CLP}$  system over  $\mathcal{OSF}$  constraints and functions over them (rewrite rules); namely:

$$\mathcal{LIFE} = \mathcal{CLP}(\mathcal{OSF} + \mathcal{FP})$$

#### LIFE—Rules + constraints for Semantic Web reasoning



#### The same hierarchy in Java

```
interface adultPerson {
  name id;
  date dob;
  int age;
  String ssn;
interface employee extends adultPerson {
  title position;
  String institution;
  employee supervisor;
  int salary;
interface marriedPerson extends adultPerson {
  marriedPerson spouse;
interface marriedEmployee extends employee, marriedPerson {
interface richEmployee extends employee {
```

#### The same hierarchy in $\mathcal{LIFE}$

```
employee <: adultPerson.</pre>
marriedPerson <: adultPerson.
richEmployee <: employee.</pre>
marriedEmployee <: employee.</pre>
marriedEmployee <: marriedPerson.</pre>
:: adultPerson
                      (id \Rightarrow name)
                      , dob \Rightarrow date
                      , age \Rightarrow int
                      , ssn \Rightarrow string).
                      (position \Rightarrow title
:: employee
                      , institution \Rightarrow string
                      , supervisor \Rightarrow employee
                      , salary \Rightarrow int).
:: marriedPerson ( spouse \Rightarrow marriedPerson ).
```

#### A relationally and functionally constrained LIFE sort hierarchy

```
:: P : adultPerson (id <math>\Rightarrow name)
                        , dob \Rightarrow date
                        , age \Rightarrow A: int
                        , ssn \Rightarrow string)
   A = ageInYears(P), A \ge 18.
                       ( position \Rightarrow T: title
:: employee
                        , institution \Rightarrow string
                        , supervisor \Rightarrow E : employee
                        , salary \Rightarrow S: int )
    higherRank(E.position, T), E.salary \geq S.
```

#### A relationally and functionally constrained LIFE sort hierarchy

*OSF* constraints as syntactic variants of logical formulae:

Sorts are unary predicates:  $X: s \iff [s]([X])$ 

Features are unary functions:  $X.f \doteq Y \iff [\![f]\!]([\![X]\!]) = [\![Y]\!]$ 

Coreferences are equations:  $X \doteq Y \iff [X] = [Y]$ 

So ...

Why not use (good old) logic proofs instead?

#### But:

model equivalence \( \neq \) proof equivalence!

- ► OSF-unification proves sort constraints by reducing them monotonically w.r.t. the sort ordering
- ▶ *ergo*, once X:s has been proven, the proof of s(X) is recorded as *the sort "s" itself!*
- ightharpoonup if further down a proof, it is again needed to prove X:s, it is remembered as X's binding
- ▶ Indeed, *OSF* constraint proof rules ensure that:

no type constraint is ever proved twice

OSF type constraints are incrementally "memoized" as they are verified:

sorts act as (instantaneous!) proof caches!

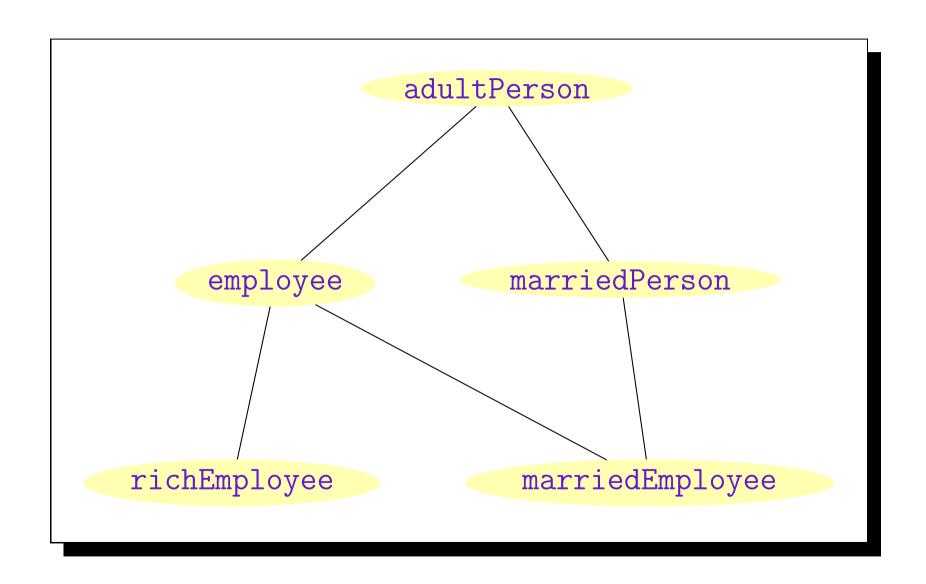
whereas in logic having proven s(X) is not "remembered" in any way (e.g., Prolog)

**Example**: consider the OSF constraint conjunction:

- X: adultPerson(age  $\Rightarrow$  25),
- $\bullet X$ : employee,
- $\bullet X : \mathtt{marriedPerson}(\mathtt{spouse} \Rightarrow Y).$

**Notation:** type#(condition) means "constraint condition attached to sort type"

#### Proof "memoizing"—Example hierarchy reminded



```
1. proving: X: adultPerson(age \Rightarrow 25)...
2. proving: adultPerson\#(X.age \ge 18) ...
3. proving: X: employee ...
4. proving: employee#(higherRank(E.position, P))...
5. proving: employee#(E.salary \geq S)...
6. proving: X: marriedPerson(spouse \Rightarrow Y)...
7. proving: X : marriedEmployee(spouse \Rightarrow Y) \dots
8. proving: marriedEmployee\#(Y.spouse = X) \dots
```

Therefore, all other inherited conditions coming from a sort greater than marriedEmployee (such as employee or adultPerson) can be safely ignored!

This "memoizing" property of OSF constraint-solving enables:

## using rules over ontologies

as well as, conversely,

## enhancing ontologies with rules

Indeed, with OSF:

- concept ontologies may be used as constraints by rules for inference and computation
- rule-based conditions in concept definitions may be used to magnify expressivity of ontologies thanks to the proof-memoizing property of ordered sorts

#### **Reasoning and the Semantic Web**

#### Outline

- **►** Constraint Logic Programming
- ▶ What is unification?
- Semantic Web objects
- Graphs as constraints
- $ightharpoonup \mathcal{OWL}$  and  $\mathcal{DL}$ -based reasoning
- Constraint-based Semantic Web reasoning
- Recapitulation

#### Recapitulation—what you must remember from this talk...

- Objects are graphs
- Graphs are constraints
- Constraints are good: they provide both formal theory and efficient processing
- ► Formal Logic is not all there is
- ▶ even so: model theory ≠ proof theory
- indeed, due to its youth, much of W3C technology is often naïve in conception and design
  - Ergo... it is condemned to reinventing [square!] wheels as long as it does not realize that such issues have been studied in depth for the past 50 years in theoretical CS!

#### Recapitulation—what you must remember from this talk...(ctd)

Pending issues re. "ontological programming"

- ► Syntax:
  - What's **essential**?
  - What's superfluous?

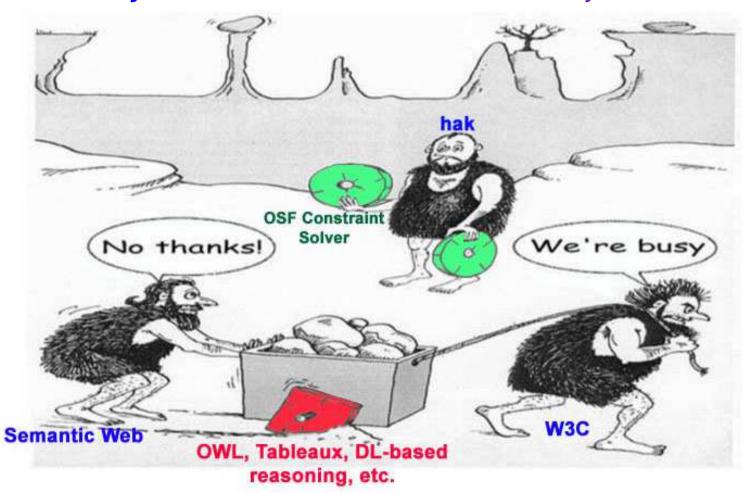
**Confusing notation**: XML-based cluttered verbosity *ok, not for human consumption—but still!* 

- Semantics:
  - What's a *model* good for?
  - What's (efficiently) provable?
  - decidable ≠ efficient
  - undecidable ≠ inefficient
- ► Applications, maintenance, evolution, etc., ...
- ► *Many, many, publications*... but no (real) field testing as yet!

## Proposal: take heed of the following facts:

- ► Linked data represents all information as interconnected sorted labelled RDF graphs—it has become a universal de facto knowledge model standard
- ▶ Differences between  $\mathcal{DL}$  and  $\mathcal{OSF}$  can come handy:
  - $-\mathcal{DL}$  is expansive—therefore, expensive—and can only describe finitely computable sets; whereas,
  - OSF is contractive—therefore, efficient—and can also describe recursively-enumerable sets
- $ightharpoonup \mathcal{CLP}$ -based graph unification reasoning = practical KR:
  - **structural**: objects, classes, inheritance
  - non-structural: path equations, relational constraints, type definitions

# If I'd asked my customers what they wanted, they'd have said a faster horse!—Henry Ford



## Thank You For Your Attention!

#### For more information:

hak@acm.org

http://cs.brown.edu/people/pvh/CPL/Papers/v1/hak.pdf

http://cedar.liris.cnrs.fr

