# Transparencies for the course TDDA41 Logic Programming, given at the Department of Computer and Information Science, Linköping University, Sweden. 

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## Introduction: Overview

- Goals of the course.
- What is logic programming?
- Why logic programming?


## Goals of the course

- Logic as a specification AND programming language;
- Theoretical foundation of logic programming;
- Practice of Prolog and constraint programming;
- Relations to other areas:
- Databases
- Formal/natural languages
- Combinatorial problems
- To program DECLARATIVELY.


## Declarative vs imperative languages

|  | Imperative | Declarative |
| :--- | :--- | :--- |
| Paradigm | Describe HOW <br> TO solve the <br> problem | Describe <br> WHAT the <br> problem is |
| Program | A sequence of <br> commands | A set of state- <br> ments |
| Examples | C, Fortran, |  |
| Ada, Java | Prolog, Pure <br> Lisp, Haskell, <br> ML |  |
| Advantages | Fast, special- <br> ized programs | General, <br> readable, <br> correct(?) <br> programs. |

Declarative description A grandchild to $x$ is a child of one of $x$ 's children.

Imperative description $I$ To find a grandchild of $x$, first find a child of $x$. Then find a child of that child.

Imperative description II To find a grandchild of $\times$, first find a parent-child pair and then check if the parent is a child of $x$.

Imperative description III To find a grandchild of $x$, compute the factorial of 123 , then find a child of $x$. Then find a child of that child.

## Compare ...

```
read(person);
for i := 1 to maxparent do
    if parent[i;1] = person then
        for \(j\) := 1 to maxparent do
            if parent[j;1] = parent[i;2] then
                write(parent[j;2]);
            fi
        od
            fi
od
```

with . . . $\mathrm{gc}(\mathrm{X}, \mathrm{Z}):-\mathrm{c}(\mathrm{X}, \mathrm{Y}), \mathrm{c}(\mathrm{Y}, \mathrm{Z})$.

## Logic: Overview

- Syntax and semantics
- Vocabulary, terms and formulas
- Interpretations and models
- Logical consequence and equivalence
- Proofs/derivations
- Soundness and completeness


# Predicate logic vocabulary 

- Constants (17, george, $t E X, \ldots$ )
- Functors (cons $/ 2,+/ 2$, father $/ 1, \ldots$ )
- Predicate symbols
(member/2, $</ 2$, father $/ 1, \ldots$ )
- Variables ( $X, X 11, \ldots, \_123, T e X, \ldots$ )
- Logical connectives ( $\wedge, \vee, \supset, \neg, \leftrightarrow)$
- Quantifiers $(\forall, \exists)$
- Auxiliary symbols (., (, ),...)


## Example

$$
A=\{\text { volvo; owner } / 1 ; \text { owns } / 2, \text { happy } / 1\}
$$

## Terms

Let A be a vocabulary.

The set of all terms over A is the least set such that

- every constant in A is a term;
- every variable is a term;
- if $f / n$ is a functor in A and $t_{1}, \ldots, t_{n}$ are terms over $\mathbf{A}$ then $f\left(t_{1}, \ldots, t_{n}\right)$ is a term.


## Ground terms

## A term that contains no variables is called a ground term.

## (Well-formed) formulas

Let A be a vocabulary.

The set of all formulas over A is the least set such that:

- if $p / n$ is a predicate symbol in A and $t_{1}, \ldots, t_{n}$ are terms, then $p\left(t_{1}, \ldots, t_{n}\right)$ is a formula;
- if $F$ and $G$ are formulas, then $(F \wedge G),(F \vee G),(F \supset G),(F \leftrightarrow G)$ and $\neg F$ are formulas;
- if $F$ is a formula and $X$ a variable, then $\forall X F$ and $\exists X F$ are formulas.


## Atoms

A formula of the form $p\left(t_{1}, \ldots, t_{n}\right)$ is called an atomic formula (atom).

## Free occurrences of variables

An occurrence of $X$ in a formula is said to be free iff the occurrence does not follow immediately after a quantifier, or in a formula immediately after $\forall X$ or $\exists X$.

## Closed formulas

A formula that does not contain any free occurrences of variables is said to be closed.

## Universal closure

Assume that $\left\{X_{1}, \ldots, X_{n}\right\}$ are the only free occurrences of variables in a formula $F$. The universal closure $\forall F$ of $F$ is the closed formula $\forall X_{1} \ldots \forall X_{n} F$.

The existential closure $\exists F$ is defined similarly.

## Interpretations

Let A be a vocabulary.

An interpretation $\Im$ of $A$ consists of (1) a non-empty set $D$ (often written $|\Im|$ ) of objects (the domain of $\Im$ ) and (2) a function that maps:

- every constant $c$ in $A$ on an element $c_{\Im}$ in D;
- every functor $f / n$ in A on a function $f_{\Im}: D^{n} \rightarrow D$;
- every predicate symbol $p / n$ in A on a relation $p_{\Im} \subseteq D^{n}$.


## Example

The vocabulary:

$$
A=\{\text { volvo; owner } / 1 ; \text { owns } / 2, \text { happy } / 1\}
$$

Consider $\Im$ where $|\Im|=\{0,1,2, \ldots\}$ and were:

- $\operatorname{volvors}^{\text {s }}=0$
- owner $_{\Im}(x)=x+1$
- owns ${ }_{\Im}=$ greater-than
- happy $_{\Im}=$ nonzero-property


## NOTE!

An interpretation defines how to interpret constants, functors and predicate symbols but it does not say what a variable denotes.

## Valuation

A valuation is a function from variables to objects in the domain of an interpretation.

## The interpretation of terms

Let $\Im$ be an interpretation of a vocabulary A. Let $\sigma$ be a valuation.

The interpretation $\sigma_{\Im}(t)$ of the term $t$ is an object in $\Im ' s$ domain:

- if $t$ is a constant $c$ then $\sigma_{\Im}(t)=c_{\Im}$;
- if $t$ is a variable $X$ then $\sigma_{\Im}(t)=\sigma(X)$;
- if $t$ is a term $f\left(t_{1}, \ldots, t_{n}\right)$ then $\sigma_{\Im}(t)=f_{\Im}\left(\sigma_{\Im}\left(t_{1}\right), \ldots, \sigma_{\Im}\left(t_{n}\right)\right)$.


## Example

Consider $\Im$ where $|\Im|=\{0,1,2, \ldots\}$ and were:

- $\operatorname{volvos}_{\Im}=0$
- owner $_{\Im}(x)=x+1$

Then:

$$
\begin{aligned}
& \left.\sigma_{\Im}\left(\operatorname{owner}^{(o w n e r}(\mathrm{volvo})\right)\right) \\
= & \operatorname{owner}_{\Im}\left(\sigma_{\Im}(\mathrm{owner}(\mathrm{volvo}))\right) \\
= & \left(\sigma _ { \Im } \left(\operatorname{owner}^{(\mathrm{volvo})))+1}\right.\right. \\
= & \left(\operatorname{owner}_{\Im}\left(\sigma_{\Im}(\mathrm{volvo})\right)\right)+1 \\
= & \left(\left(\sigma_{\Im}\left(\operatorname{volvo}_{\mathrm{vo}}\right)\right)+1\right)+1 \\
= & \left(\left(\operatorname{volvo}_{\Im}\right)+1\right)+1 \\
= & (0+1)+1 \\
= & 2
\end{aligned}
$$

## Example

Consider also $\sigma(\mathrm{X})=3$. Then:

$$
\begin{aligned}
& \sigma_{\Im}(\operatorname{owner}(\mathrm{X})) \\
= & \text { owner }_{\Im}\left(\sigma_{\Im}(\mathrm{X})\right) \\
= & \left(\sigma_{\Im}(\mathrm{X})\right)+1 \\
= & (\sigma(\mathrm{X}))+1 \\
= & 3+1 \\
= & 4
\end{aligned}
$$

## The interpretation of formulas

The meaning of a formula is a truth-value-"true" or "false". Given an interpretation $\Im$ and a valuation $\sigma$ we write
$\Im \models_{\sigma} F$ when $F$ is true wrt $\Im$ and $\sigma$.
$\Im \not \vDash_{\sigma} F$ when $F$ is false wrt $\Im$ and $\sigma$.

- $\Im \models_{\sigma} p\left(t_{1}, \ldots, t_{n}\right)$ iff
$\left(\sigma_{\Im}\left(t_{1}\right), \ldots, \sigma_{\Im}\left(t_{n}\right)\right) \in p_{\Im} ;$
- $\Im \models_{\sigma} \neg F$ iff $\Im \not \models_{\sigma} F$;
- $\Im \models_{\sigma} F \wedge G$ iff $\Im \models_{\sigma} F$ and $\Im \models_{\sigma} G$;
- $\Im \models_{\sigma} F \vee G$ iff $\Im \models_{\sigma} F$ and/or $\Im \models_{\sigma} G$;


## The interpretation of formulas (cont'd.)

- $\Im \models_{\sigma} F \supset G$ iff $\Im \not \vDash_{\sigma} F$ and/or $\Im \models_{\sigma} G$;
- $\Im \models_{\sigma} F \leftrightarrow G$ iff $\Im \models_{\sigma} F$ exactly when $\Im \vDash{ }_{\sigma} G$;
- $\Im \models_{\sigma} \forall X F$ iff $\Im \models_{\sigma[x \mapsto t]} F$ for every $t \in|\Im| ;$
- $\Im \neq{ }_{\sigma} \exists X F$ iff $\Im \models_{\sigma[x \mapsto t]} F$ for some $t \in|\Im|$.


## Example

Consider $\Im$ as before.
Then:

$$
\Im \models \text { owns(volvo, volvo) } \supset \text { happy(volvo) }
$$

iff

$$
\begin{aligned}
& \Im \not \models \text { owns(volvo, volvo) } \\
& \text { or } \\
& \Im \models \text { happy(volvo) }
\end{aligned}
$$

iff

$$
\begin{aligned}
& \left\langle\sigma_{\Im}(\text { volvo }), \sigma_{\Im}(\text { volvo })\right\rangle \notin \text { owns }_{\Im} \\
& \text { or } \\
& \sigma_{\Im}(\text { volvo }) \in \text { happy }_{\Im}
\end{aligned}
$$

iff

$$
\langle 0,0\rangle \notin \text { owns }_{\Im} \text { or } 0 \in \text { happy }_{\Im}
$$

iff
$0 \ngtr 0$ or $0 \neq 0$
iff
true

## Models

Let $F$ be a closed formula.
Let $P$ be a set of closed formulas.

An interpretation $\Im$ is a model of $F$ iff $\Im \models F$.

An interpretation $\Im$ is a model of $P$ iff $\Im$ is a model of every formula in $P$.

## Satisfiability

$F$ (resp. $P$ ) is satisfiable iff $F$ (resp. $P$ ) have at least one model. (Otherwise $F / P$ is unsatisfiable.)

## Example

$\Im$ (defined as before) is a model of: owns(owner(volvo), volvo)
and:

$$
\forall X(o w n s(X, \text { volvo) } \supset \text { happy }(X))
$$

## Logical consequence

$F$ is a logical consequence of $P(P \models F)$ iff $F$ is true in all of $P$ 's models
$(\operatorname{Mod}(P) \subseteq \operatorname{Mod}(F))$.

## Theorem

$P \vDash F$ iff $P \cup\{\neg F\}$ is unsatisfiable.

## Logical equivalence

Let $F, G, \forall X H(X)$ be formulas.
$F$ and $G$ are logically equivalent ( $F \equiv G$ ) iff $\Im \models_{\sigma} F$ exactly when $\Im \models_{\sigma} G$.

$$
\begin{aligned}
F \supset G & \equiv \neg F \vee G \\
F \supset G & \equiv \neg G \supset \neg F \\
F \leftrightarrow G & \equiv(F \supset G) \wedge(G \supset F) \\
\neg(F \wedge G) & \equiv \neg F \vee \neg G \\
\neg(F \vee G) & \equiv \neg F \wedge \neg G \\
\neg \forall X H(X) & \equiv \exists X \neg H(X) \\
\neg \exists X H(X) & \equiv \forall X \neg H(X)
\end{aligned}
$$

In addition, if $X$ does not occur free in $F$.

$$
\forall X(F \vee H(X)) \equiv F \vee \forall X H(X)
$$

## Proofs (derivations)

A proof (derivation) is a sequence of formulas where each formula in the sequence is either a so-called premise or is obtained from previous formulas in the sequence by means of a collection of derivation rules.

## Natural deductions



## Example

1. owns(owner(volvo), volvo) $P$
2. $\forall \mathrm{X}(\mathrm{owns}(\mathrm{X}, \mathrm{volvo}) \supset \operatorname{happy}(\mathrm{X})) \quad P$
3. owns(owner(volvo), volvo) D happy(owner(volvo)))
4. happy(owner(volvo))

## Proofs

Let $P$ be a set of closed formulas (premises) Let $F$ be a closed formula.

We write $P \vdash F$ when there is a derivation of $F$ from the premises $P$.

Soundness and completeness

If $P \vdash F$ then $P \models F$. (soundness)

If $P \vDash F$ then $P \vdash F$. (completeness)

# Definite Programs: Overview 

- Definite programs:
- Rules;
- Facts;
- Goals.
- Herbrand-interpretations;
- Herbrand-models;
- Fixpoint-semantics.


## Clauses

A clause is a formula:

$$
\forall\left(A_{1} \vee \ldots \vee A_{m} \vee \neg A_{m+1} \vee \ldots \vee \neg A_{m+n}\right)
$$

where $A_{1}, \ldots, A_{m}, A_{m+1}, \ldots, A_{m+n}$ are atoms and $m, n \geq 0$.

$$
\begin{aligned}
\forall\left(A_{1} \vee \ldots \vee A_{m} \vee\right. & \left.\neg A_{m+1} \vee \ldots \vee \neg A_{m+n}\right) \\
& \equiv \\
\forall\left(\left(A_{1} \vee \ldots \vee A_{m}\right) \vee\right. & \left.\neg\left(A_{m+1} \wedge \ldots \wedge A_{m+n}\right)\right) \\
& \equiv \\
\forall\left(\left(A_{1} \vee \ldots \vee A_{m}\right)\right. & \left.\leftarrow\left(A_{m+1} \wedge \ldots \wedge A_{m+n}\right)\right)
\end{aligned}
$$

## Definite clauses

A definite clause is a clause where $m \leq 1$ :

Rules

A rule is a clause where $m=1$ and $n>0$ :

$$
\forall\left(A_{1} \leftarrow A_{2} \wedge \ldots \wedge A_{m+n}\right)
$$

Facts

A fact is a clause where $m=1$ and $n=0$ :
$\forall\left(A_{1}\right)$

## (Definite) goals

A goal is a clause where $m=0$ and $n \geq 0$ :

$$
\forall\left(\neg\left(A_{1} \wedge \ldots \wedge A_{m+n}\right)\right)
$$

A goal where $m=n=0$ is called the empty goal.

## Notation

Rules: $\quad A_{1} \leftarrow A_{2}, \ldots, A_{n+1} . \quad n>0$
Facts: $A_{1}$.
Goals: $\leftarrow A_{1}, \ldots, A_{n}$.

$$
\begin{aligned}
& n>0 \\
& n=0
\end{aligned}
$$

# Logic Programming Anatomy 

$$
\begin{array}{ccc}
\text { head } & \text { neck } & \text { body } \\
A_{0} & \leftarrow & A_{1}, \ldots, A_{n}
\end{array}
$$

## Logic programs

A definite program is a finite set of rules and facts.

A definite program $P$ is used to answer
"existential questions" (queries) such as:
"are there any odd integers?"

The query can be answered "yes" if e.g:

$$
P \models \exists X \operatorname{odd}(X)
$$

This is equivalent to proving that:

$$
P \cup\{\neg \exists X \operatorname{odd}(X)\}
$$

is unsatisfiable (has no models).

## Resolution

Note that $\neg \exists\left(A_{1} \wedge \ldots \wedge A_{n}\right)$ is equivalent to $\forall \neg\left(A_{1} \wedge \ldots \wedge A_{n}\right)$. That is, a goal.

Resolution is used to prove that a set of clauses is unsatisfiable. As a side-effect resolution produces "witnesses" (variable bindings). See chapter 3.

## Herbrand interpretations

Let $P$ be a logic program based on the vocabulary $A$

## Herbrand universe

The Herbrand universe of $P$ ( $A$ really) is the set of all ground terms that can be built using constants and functors in $P(A)$. Denoted $U_{P}\left(U_{A}\right)$.

## Herbrand base

The Herbrand base of $P(A)$ is the set of all ground atoms that can be built using $U_{P}$ and the predicate symbols of $P(A)$. Denoted $B_{P}$ $\left(B_{A}\right)$.

## Example

Vocabulary:

$$
A=\{\text { volvo; owner } / 1 ; \text { owns } / 2, \text { happy } / 1\}
$$

Herbrand universe:

$$
U_{A}=\{\text { volvo, owner(volvo), owner(owner(volvo)), ... }\}
$$

Herbrand base:

$$
B_{A}=\left\{\operatorname{happy}(s) \mid s \in U_{A}\right\} \cup\left\{o w n s(s, t) \mid s, t \in U_{A}\right\}
$$

## Herbrand interpretations

A Herbrand interpretation of $P$ is an interpretation $\Im$ where $|\Im|=U_{P}$ and where:

- $c_{\Im}=c$ for every constant $c$;
- $f_{\Im}\left(t_{1}, \ldots, t_{n}\right)=f\left(t_{1}, \ldots, t_{n}\right)$ for every functor $f / n$;
- $p_{\Im}$ is a subset of $\underbrace{U_{P} \times \cdots \times U_{P}}_{n}$ for every predicate symbol $p / n$.

That is, the interpretation of a ground term is the term itself!

## Observation I

Since all ground terms are interpreted as themselves, it is sufficient to specify the interpretation of the predicate symbols when describing a Herbrand interpretation; in other words, to specify a Herbrand interpretation $\Im$ it is sufficient to specify, for each predicate symbol, the set:

$$
\left\{\left\langle t_{1}, \ldots, t_{n}\right\rangle \in U_{P}^{n} \mid p\left(t_{1}, \ldots, t_{n}\right) \text { is true in } \Im\right\}
$$

## Observation II

Instead of describing a Herbrand interpretation $\Im$ as a family of sets we usually describe $\Im$ as a single set of all ground atoms that are true in $\Im$.

$$
\Im=\left\{p\left(t_{1}, \ldots, t_{n}\right) \mid p\left(t_{1}, \ldots, t_{n}\right) \text { is true in } \Im\right\}
$$

## Example

## Alternative I

$$
\begin{aligned}
\text { owns }_{\Im} & =\{\langle\text { owner(volvo) }, \text { volvo }\rangle, \ldots\} \\
\text { happy }_{\Im} & =\{\langle\text { owner(volvo) }\rangle, \ldots\}
\end{aligned}
$$

Alternative II

$$
\begin{aligned}
\Im= & \{o w n s(o w n e r(\text { volvo }), \text { volvo }), \ldots, \\
& \text { happy(owner(volvo)) }, \ldots\}
\end{aligned}
$$

## Ground instances of $P$

Let $C$ be a definite clause of the form

$$
A_{0} \leftarrow A_{1}, \ldots, A_{n} \quad(n \geq 0)
$$

( $C$ is considered to be a fact if $n=0$.)

By a ground instance of $C$ we mean the same clause with all variables replaced by ground terms (several occurrences of the same variable are replaced by the same term):

By ground (C) we mean the set of all ground instances of $C$.

If $P$ is a definite program then
$\operatorname{ground}(P)=\left\{C^{\prime} \mid \exists C \in P\right.$ s.t. $\left.C^{\prime} \in \operatorname{ground}(C)\right\}$

## Why Herbrand Interpretations?

For an arbitrary interpretation $\Im$ :

$$
\begin{gathered}
\Im \models_{\sigma} \forall X(\operatorname{happy}(X) \leftarrow \operatorname{iff}) \\
\Im \operatorname{lowns}(X, \text { volvo })) \\
{ }_{\sigma[X \mapsto a]} \operatorname{happy}(X) \leftarrow \operatorname{owns}(X, \text { volvo }) \\
\text { for all } a \in|\Im|
\end{gathered}
$$

For a Herbrand interpretation $\Im$ :

$$
\begin{gathered}
\Im \models_{\sigma} \forall X(\operatorname{happy}(X) \leftarrow \operatorname{iff}) \\
\left.\Im \operatorname{lowns}^{( }(X, \text { volvo })\right) \\
\text { happy }(t) \leftarrow \text { owns }(t, \text { volvo }) \\
\text { for any } t \in U_{P}
\end{gathered}
$$

No need to worry about valuations!!!

## Herbrand models

A Herbrand model of $F$ (resp. $P$ ) is a Herbrand interpretation which is a model of $F$ (resp. all formulas in $P$ ).

## Observation

A ground atom $A$ is true in a Herbrand interpretation $\Im$ iff $A \in \Im$.

## Theorem

Let $P$ be a set of definite clauses
(facts/rules/goals) and $M$ be an arbitrary model of $P$. Then:

$$
\Im:=\left\{A \in B_{P} \mid M \models A\right\}
$$

is a Herbrand model of $P$.

## Theorem

Let $\left\{M_{1}, M_{2}, \ldots\right\}$ be a non-empty set of Herbrand models of $P$. Then also $\Im:=\cap\left\{M_{1}, M_{2}, \ldots\right\}$ is a Herbrand model of $P$.

## The Least Herbrand model

The intersection of all Herbrand models of $P$ is called the least Herbrand model of $P$ and is denoted $M_{P}$.

Theorem

$$
M_{P}=\left\{A \in B_{P}|P|=A\right\}
$$

## "Construction" of $M_{P}$

Observation
In order for $\Im$ to be a model of $P$ it is required that:

- If $A$ is a ground instance of a fact then $A \in \Im$, and
- If $A \leftarrow A_{1}, \ldots, A_{n}$ is a ground instance of a clause in $P$ and $\left\{A_{1}, \ldots, A_{n}\right\} \subseteq \Im$ then $A \in \Im$.


## Immediate consequence operator

$$
\begin{aligned}
T_{P}(x):= & \\
\left\{A \in B_{P} \quad \mid\right. & A \leftarrow A_{1}, \ldots, A_{n} \in \operatorname{ground}(P) \\
& \text { and } \left.\left\{A_{1}, \ldots, A_{n}\right\} \subseteq x\right\}
\end{aligned}
$$

## Theorem

$$
M_{P}=T_{P}^{n}(\emptyset) \quad \text { when } n \rightarrow \infty
$$

## Example

$g p(X, Y):-p(X, Z), p(Z, Y)$.
$p(X, Y):-f(X, Y)$.
$p(X, Y):-m(X, Y)$.
f(adam,bill).
f(adam, carol).
f(bill,eve).
m(carol,david).

## Example

- $\Im_{0}=\emptyset$
- $\Im_{1}=T_{P}(\emptyset)=\{f(a, b), f(a, c), f(b, e), m(c, d)\}$
$\left[f(a, b) \in \Im_{1}\right.$ since $(f(a, b) \leftarrow) \in \operatorname{ground}(P)$ and $\emptyset \subseteq \emptyset$.
- $\Im_{2}=T_{P}\left(\Im_{1}\right)=T_{P}^{2}(\emptyset)=$ $\{p(a, b), p(a, c), p(b, e), p(c, d)\} \cup \Im_{1}$
$\left[p(a, b) \in \Im_{2}\right.$ since $(p(a, b) \leftarrow f(a, b)) \in \operatorname{ground}(P)$ and $\{f(a, b)\} \subseteq \Im_{1}$.]
- $\Im_{3}=T_{P}\left(\Im_{2}\right)=T_{P}^{3}(\emptyset)=\{g p(a, d), g p(a, e)\} \cup \Im_{2}$
$\left[g p(a, d) \in \Im_{3}\right.$ since
$(g p(a, d) \leftarrow p(a, c), p(c, d)) \in \operatorname{ground}(P)$ and $\left.\{p(a, c), p(c, d)\} \subseteq \Im_{2}.\right]$
- $\Im_{4}=T_{P}\left(\Im_{3}\right)=T_{P}^{4}(\emptyset)=\Im_{3}$


# SLD-Resolution: Overview 

- Substitutions;
- Unification;
- SLD-derivations;
- Soundness and completeness.


## Substitutions

A substitution is a finite set
$\left\{X_{1} / t_{1}, \ldots, X_{n} / t_{n}\right\}$ where:

- every $t_{i}$ is a term;
- every $X_{i}$ is a variable distinct from $t_{i}$;
- if $i \neq j$ then $X_{i} \neq X_{j}$.

The empty substitution $\}$ is denoted $\epsilon$.

Let $\theta$ be a substitution $\left\{X_{1} / t_{1}, \ldots, X_{n} / t_{n}\right\}$.

## Domain and Range

The domain $\operatorname{Dom}(\theta)$ of $\theta$ is $\left\{X_{1}, \ldots, X_{n}\right\}$ and the range $\operatorname{Range}(\theta)$ is the set of all variables occurring in $t_{1}, \ldots, t_{n}$.

## Application

Let $E$ be a term or formula. The application $E \theta$ of $\theta$ to $E$ is the term/formula obtained from $E$ by simultaneously replacing all occurrences of $X_{i}$ by $t_{i}$.
$E \theta$ is called an instance of $E$.

## Composition

Let $\theta:=\left\{X_{1} / s_{1}, \ldots, X_{m} / s_{m}\right\}$ and $\sigma:=\left\{Y_{1} / t_{1}, \ldots, Y_{n} / t_{n}\right\}$ be substitutions. The composition $\theta \sigma$ of $\theta$ and $\sigma$ is the substitution obtained from

$$
\left\{X_{1} / s_{1} \sigma, \ldots, X_{m} / s_{m} \sigma, Y_{1} / t_{1}, \ldots, Y_{n} / t_{n}\right\}
$$

by removing all $X_{i} / s_{i} \sigma$ where $X_{i}=s_{i} \sigma$ and all $Y_{i} / t_{i}$ where $Y_{i} \in \operatorname{Dom}(\theta)$.

## More general substitution

A substitution $\theta$ is more general than $\sigma$ ( $\sigma \preceq \theta$ ) iff there exists a substitution $\omega$ such that $\theta \omega=\sigma$.

## Theorem

## Let $\theta, \sigma$ and $\gamma$ be substitutions and $E$ a term/formula. Then

- $(\theta \sigma) \gamma=\theta(\sigma \gamma)$;
- $E(\theta \sigma)=(E \theta) \sigma$;
- $\epsilon \theta=\theta \epsilon=\theta$.


## Unification

A structure is a term or an atomic formula.

## Unifier

A unifier of two structures $s$ and $t$ is a substitution $\theta$ such that $s \theta=t \theta$.

## Most general unifier (mgu)

A unifier $\theta$ of $s$ and $t$ is called a most general unifier of $s$ and $t$ iff $\sigma \preceq \theta$ for every unifier $\sigma$ of $s$ and $t$. NB: Two unifiable structures have at least one mgu (usually infinitely many).

## Solved form

A set of equation $\left\{s_{1} \doteq t_{1}, \ldots, s_{n} \doteq t_{n}\right\}$ is in solved form iff $s_{1}, \ldots, s_{n}$ are distinct variables none of which occur in $t_{1}, \ldots, t_{n}$.

## Solution

A substitution $\theta$ is a solution to a set of equations $\left\{s_{1} \doteq t_{1}, \ldots, s_{n} \doteq t_{n}\right\}$ iff $\theta$ is a unifier of $s_{i}$ and $t_{i}(1 \leq i \leq n)$.

## Theorem

If $\left\{X_{1} \doteq t_{1}, \ldots, X_{n} \doteq t_{n}\right\}$ is in solved form then $\left\{X_{1} / t_{1}, \ldots, X_{n} / t_{n}\right\}$ is an mgu of $X_{i}$ and $t_{i}(1 \leq i \leq n)$.
select an arbitrary $s \doteq t \in E$;
case $s \doteq t$ of
$f\left(s_{1}, \ldots, s_{n}\right) \doteq f\left(t_{1}, \ldots, t_{n}\right)$
where $n \geq 0 \Rightarrow$
replace equation by $s_{1} \doteq t_{1}, \ldots, s_{n} \doteq t_{n}$;
$f\left(s_{1}, \ldots, s_{m}\right) \doteq g\left(t_{1}, \ldots, t_{n}\right)$
where $f / m \neq g / n \Rightarrow$
halt with $\perp$;
$X \doteq X \Rightarrow$
remove the equation;
$t \doteq X$ where $t$ is not a variable $\Rightarrow$
replace equation by $X \doteq t$;
$X \doteq t$ where $X \neq t$ and $X$ has more than
one occurrence in $E \Rightarrow$
if $X$ is a proper subterm of $t$ then halt with $\perp$
else

$$
\begin{aligned}
& \text { replace all other occurrences } \\
& \text { of } X \text { by } t \text {; }
\end{aligned}
$$

esac

## Theorem

The algorithm always terminates. If $s$ and $t$ are unifiable then the algorithm returns a solved form whose mgu is an mgu of $s$ and $t$. Otherwise the algorithm returns $\perp$.

Renaming

A substitution $\theta:=\left\{X_{1} / Y_{1}, \ldots, X_{n} / Y_{n}\right\}$ where $Y_{1}, \ldots, Y_{n}$ is a permutation of $X_{1}, \ldots, X_{n}$ is called a renaming. The substitution
$\left\{Y_{1} / X_{1}, \ldots, Y_{n} / X_{n}\right\}$ is called the inverse of $\theta$ (denoted $\theta^{-1}$ ).

## Theorem

Let $\theta$ and $\sigma$ be mgu's of $s$ and $t$. Then there exists a renaming $\gamma$ such that $\theta \gamma=\sigma$ (and $\sigma \gamma^{-1}=\theta$ ).

## Theorem

If $\theta$ is an mgu of $s$ and $t$ and $\sigma$ a renaming, then $\theta \sigma$ is also an mgu of $s$ and $t$.

## In practice

The previous algorithm is worst-case exponential in the size of the structures.
Take for instance
$g\left(X_{1}, \ldots, X_{n}\right)=g\left(f\left(X_{0}, X_{0}\right), \ldots, f\left(X_{n-1}, X_{n-1}\right)\right)$.
The reason is the occurs check (i.e. checking if $X$ is a proper subterm of $t$ ).

There are also polynomial algorithms, but most Prolog implementations use the exponential algorithm, and simply drop the occurs check.

This rarely makes a difference, but does make Prolog unsound!!!

## SLD-resolution rule

Let $H \leftarrow B_{1}, \ldots, B_{n}$ be a program clause renamed apart from $\leftarrow A_{1}, \ldots, A_{i}, \ldots, A_{m}$, and let $\theta$ be an mgu of $A_{i}$ and $H$. Then:

$$
\frac{\leftarrow A_{1}, \ldots, A_{i}, \ldots, A_{m} \quad H \leftarrow B_{1}, \ldots, B_{n}}{\leftarrow\left(A_{1}, \ldots, A_{i-1}, B_{1}, \ldots, B_{n}, A_{i+1}, \ldots, A_{m}\right) \theta}
$$

## SLD-derivation

Let $G_{0}$ be a goal. An SLD-derivation of $G_{0}$ is a finite/infinite sequence:

$$
G_{0} \stackrel{C_{0}}{\sim} G_{1} \cdots G_{n-1} \stackrel{C_{n-1}}{\sim} G_{n} \cdots
$$

of goals and (renamed) program clauses such that:

$$
\frac{G_{i} \quad C_{i}}{G_{i+1}}
$$

$g p(X, Y):-p(X, Z), p(Z, Y)$.
$p(X, Y):-f(X, Y)$.
$p(X, Y):-m(X, Y)$.
f(adam,tom).
f(adam, mary).
f(tom, david).
m(mary, anne).

$$
\begin{aligned}
& \operatorname{inv}(0,1) . \\
& \operatorname{inv}(1,0) .
\end{aligned}
$$

```
and(0,0,0).
and (0,1,0).
and(1,0,0).
and (1, 1, 1).
```

$\operatorname{nand}(X, Y, Z):-\operatorname{and}(X, Y, W), \operatorname{inv}(W, Z)$.

## Computation rule

A computation rule $\Re$ is a (partial) function that given a goal returns an atom in that goal.

SLD-refutation

An SLD-refutation of $G_{0}$ is a finite SLD-derivation

$$
G_{0} \stackrel{C_{0}}{\sim} G_{1} \cdots G_{n-1} \stackrel{C_{n-1}}{\sim} G_{n}
$$

where $G_{n}=\square$.

## Failed derivation

A finite SLD-derivation

$$
G_{0} \stackrel{C_{0}}{\sim} G_{1} \cdots G_{n-1} \stackrel{C_{n-1}}{\sim} G_{n}
$$

is said to be failed if the selected atom in $G_{n}$ does not unify with any program clause head.

## Complete SLD-derivation

An SLD-derivation is complete if it is a refutation, a failed or infinite derivation.

Let

$$
G_{0} \stackrel{C_{0}}{\sim} G_{1} \cdots G_{n-1} \stackrel{C_{n-1}}{\sim} G_{n}
$$

be an SLD-derivation

## Computed substitution

If $\theta_{i}$ is mgu $i$ of the derivation then

$$
\theta_{1} \theta_{2} \ldots \theta_{n}
$$

is called the computed substitution in the derivation.

## Computed answer-substitution

The computed answer-substitution in a refutation of $G_{0}$ is the computed substitution of the refutation restricted to the variables occurring in $G_{0}$.

Let $P$ be a logic program;
Let $\Re$ be a computation rule

SLD-tree

The SLD-tree of a goal $G_{0}$ is a tree where

- the root of the tree is $G_{0}$;
- if $G_{i}$ is a node in the tree then $G_{i}$ has a child $G_{i+1}$ (connected via a branch labelled " $C_{i}$ ") iff there exists an SLD-derivation

$$
G_{0} \stackrel{C_{0}}{\sim} G_{1} \cdots G_{i} \stackrel{C_{i}}{\sim} G_{i+1}
$$

with the computation rule $\Re$.

## Soundness and completeness

## Theorem (soundness)

Let $P$ be a logic program, $\Re$ a computation rule and $\theta$ an $\Re$-computed answer-substitution of the goal $\leftarrow A_{1}, \ldots, A_{n}$. Then $\forall\left(\left(A_{1} \wedge \ldots \wedge A_{n}\right) \theta\right)$ is a logical consequence of $P$.

## Theorem (completeness)

Let $P$ be a logic program and $\Re$ a computation rule. If $\forall\left(A_{1} \wedge \ldots \wedge A_{n}\right) \sigma$ is a logical consequence of $P$ then there is a refutation of $\leftarrow A_{1}, \ldots, A_{n}$ with $\Re$-computed answer-substitution $\theta$ such that $\left(A_{1} \wedge \ldots \wedge A_{n}\right) \sigma$ is an instance of $\left(A_{1} \wedge \ldots \wedge A_{n}\right) \theta$.

## Example

$\%$ leq( $X, Y$ ) - $X$ is less than or equal to $Y$ leq( $0, Y$ ).
$\operatorname{leq}(s(X), s(Y)):-\operatorname{leq}(X, Y)$.
:- leq(0, N).
yes

That is $P \models \forall N \operatorname{leq}(0, N)$.

Note that it is impossible to obtain e.g. the answer $N=s(0))$. However, we get a more general answer.

# Negation: Overview 

- Closed World Assumption;
- Negation as Failure;
- Completion;
- SLDNF-resolution (part I);
- General (alt. normal) logic programs;
- Stratified logic programs;
- SLDNF-resolution (part II).


## Program:

> parent $(a, b)$.
> parent $(a, c)$.
> parent $(c, d)$.
female(a).
female(d).
mother(X) :- parent(X,Y), female(X).

Least Herbrand model:

parent (a,b).<br>parent (a, c).<br>parent(c,d).<br>female(a).<br>female(d).<br>mother (a).

## Program:

> edge $(a, b)$.
> edge $(a, c)$.
> edge $(b, d)$.
> edge,$d)$.
path $(X, Y)$ :- edge(X,Y).
path(X,Y) :- edge(X,Z), path(Z,Y).

Least Herbrand model:

$$
\begin{aligned}
& \operatorname{edge}(a, b) . \\
& \operatorname{edge}(a, c) . \\
& \operatorname{edge}(b, d) . \\
& \operatorname{edge}(c, d) . \\
& \operatorname{path}(a, b) . \\
& \operatorname{path}(a, c) . \\
& \operatorname{path}(b, d) . \\
& \operatorname{path}(c, d) . \\
& \operatorname{path}(a, d) .
\end{aligned}
$$

## Closed World Assumption

Background Definite programs can only be used to describe positive knowledge; it is not possible to describe objects that are not related.

Solution I Closed world assumption:

$$
\frac{P \not \vDash A}{\neg A}
$$

Problem $P \not \vDash A$ is undecidable.

## Negation as (finite) Failure

Solution II An SLD-tree is finitely failed iff it is finite and does not contain any refutations.

Observation If $\leftarrow A$ has a finitely failed SLD-tree then $P \not \vDash A$. (Follows from the soundness and completeness of
SLD-resolution.)

## The NAF rule

$\frac{\leftarrow A \text { has a finitely failed SLD-tree }}{\neg A}$

Problem The NAF rule is not sound.

## Completion

Thesis The program contains information that is not written out explicitly. The completed program is the program obtained after addition of the missing information.

Observation $\{a \leftarrow b, a \leftarrow c\} \equiv\{a \leftarrow b \vee c\}$.
Principle An implication $a \leftarrow b$ is replaced by an equivalence $a \leftrightarrow b$.

Let $Y_{1}, \ldots, Y_{i}$ be all variables in $p\left(t_{1}, \ldots, t_{m}\right) \leftarrow A_{1}, \ldots, A_{n}$.

Step 1 Replace the clause by

$$
\begin{aligned}
& p\left(X_{1}, \ldots, X_{m}\right) \leftarrow \\
& \quad \exists Y_{1} \ldots Y_{i}\left(X_{1} \doteq t_{1}, \ldots, X_{m} \doteq t_{m}, A_{1}, \ldots, A_{n}\right)
\end{aligned}
$$

Step 2 Take all clauses

$$
\begin{gathered}
p\left(X_{1}, \ldots, X_{m}\right) \leftarrow E_{1} \\
\vdots \\
p\left(X_{1}, \ldots, X_{m}\right) \leftarrow E_{j}
\end{gathered}
$$

that define $p / m$ and replace by

$$
\begin{array}{ll}
p\left(X_{1}, \ldots, X_{m}\right) \leftarrow E_{1} \vee \ldots \vee E_{j} & (j>0) \\
p\left(X_{1}, \ldots, X_{m}\right) \leftarrow \square & (j=0)
\end{array}
$$

Step 3 Replace all implications with equivalences.

Step 4 Add the "free equality axioms":

$$
\begin{aligned}
& X \doteq X \\
& X \doteq Y \rightarrow Y \doteq X \\
& X \doteq Y \wedge Y \doteq Z \rightarrow X \doteq Z \\
& X_{1} \doteq Y_{1} \wedge \ldots \wedge X_{m} \doteq Y_{m} \rightarrow \\
& \quad \quad\left(X_{1}, \ldots, X_{m}\right) \doteq f\left(Y_{1}, \ldots, Y_{m}\right) \\
& X_{1} \doteq Y_{1} \wedge \ldots \wedge X_{m} \doteq Y_{m} \rightarrow \\
& \quad\left(p\left(X_{1}, \ldots, X_{m}\right) \rightarrow p\left(Y_{1}, \ldots, Y_{m}\right)\right) \\
& f\left(X_{1}, \ldots, X_{m}\right) \neq g\left(Y_{1}, \ldots, Y_{n}\right) \text { if } f / m \neq g / n \\
& f\left(X_{1}, \ldots, X_{m}\right) \doteq f\left(Y_{1}, \ldots, Y_{m}\right) \rightarrow \\
& \quad X_{1} \doteq Y_{1} \wedge \ldots \wedge X_{m} \doteq Y_{m} \\
& f(\ldots \ldots) \neq X
\end{aligned}
$$

Soundness of "Negation as Failure"

Theorem Let $P$ be a definite program. If
$\leftarrow A$ has a finitely failed SLD-tree then $\operatorname{comp}(P) \models \forall \neg A$.

## Completeness of "Negation as Failure"

Theorem Let $P$ be a definite program. If $\operatorname{comp}(P) \vDash \forall \neg A$ then there exists a finitely failed SLD-tree of $\leftarrow A$.

## SLDNF-resolution for definite programs

A general goal is an expression

$$
\leftarrow L_{1}, \ldots, L_{n} .
$$

where each $L_{i}$ is an atom (positive literal) or a negated atom (negative literal).

## Combine SLD-resolution and "Negation as Failure"

Given a general goal - if the selected literal is positive then the next goal is obtained in the usual way. If the selected literal is negative $(\neg A)$ and $\leftarrow A$ has a finitely failed SLD-tree then the next goal is obtained by removing $\neg A$ from the goal.

## Soundness of SLDNF

Theorem Let $P$ be a definite program and $\leftarrow L_{1}, \ldots, L_{n}$ a general goal. If $\leftarrow L_{1}, \ldots, L_{n}$ has an SLDNF-refutation with computed answer-substitution $\theta$ then $\forall\left(L_{1} \wedge \cdots \wedge L_{n}\right) \theta$ is a logical consequence of $\operatorname{comp}(P)$.

No completeness!!!

## General (or normal) programs

A general clause is a clause of the form

$$
A \leftarrow L_{1}, \ldots, L_{n} \quad(n \geq 0)
$$

where $L_{1}, \ldots, L_{n}$ are positive/negative literals.

## Completion

Completion of a general program is obtained in the same way as for definite programs.
(Negative literals are handled like positive literals.)

## Stratified programs

Problem Completion of a general program can be inconsistent (unsatisfiable).

Limitation A stratified program is a general program where "no relation is defined in terms of its own complement". That is, no predicate symbol depends on its own negation.

## Stratified programs

A general program $P$ is stratified iff there exists a partitioning $P_{1}, \ldots, P_{n}$ of $P$ such that

- if $p(\ldots) \leftarrow \ldots, q(\ldots), \ldots \in P_{i}$ then
$\operatorname{DEF}(q) \subseteq P_{1} \cup \ldots \cup P_{i}$.
- if $p(\ldots) \leftarrow \ldots, \neg q(\ldots), \ldots \in P_{i}$ then
$\operatorname{DEF}(q) \subseteq P_{1} \cup \ldots \cup P_{i-1}$.

Theorem Completion of a stratified program is always consistent.

## SLDNF-resolution for general programs

Let $P$ be a general program, $G_{0}$ a general goal and $\Re$ a computation rule. The SLDNF-forest of $G_{0}$ is the least forest (modulo renaming) such that

1. $G_{0}$ is a root of one tree.
2. if $G$ is a node and $\Re(G)=A$ then $G$ has a child $G^{\prime}$ for each clause $C$ such that $G^{\prime}$ is obtained from $G$ and $C$. If there is no such clause, $G$ has a single child $\mathbf{F F}$;
3. if $G$ is a node of the form
$\leftarrow L_{1}, \ldots, L_{i-1}, \neg A, L_{i+1}, \ldots, L_{i+j}$ and $\Re(G)=\neg A$, then

## Cont'd

- the forest contains a tree with the root $\leftarrow A$;
- if the tree with the root $\leftarrow A$ has a leaf $\square$ with the empty computed answer-substitution, then $G$ has a child FF.
- if the tree with root $\leftarrow A$ is finite and all leaves are $\mathbf{F F}$, then $G$ has a single child $\leftarrow L_{1}, \ldots, L_{i-1}, L_{i+1}, \ldots, L_{i+j}$.


## Soundness of SLDNF-resolution

Let $P$ be a general program, $\leftarrow L_{1}, \ldots, L_{n}$ a general goal and $\Re$ a computation rule. If $\theta$ is a computed answer-substitution in an SLDNF-refutation of $\leftarrow L_{1}, \ldots, L_{n}$ then $\forall\left(\left(L_{1} \wedge \ldots \wedge L_{n}\right) \theta\right)$ is a logical consequence of $\operatorname{comp}(P)$.
father (X) :parent $(X, Y)$,
$\backslash+\operatorname{mother}(X, Y)$.
disjoint ([], X).
disjoint ([X|Xs],Ys) :-
$\+$ member (X,Ys),
disjoint(Xs,Ys).

```
founding(X) :-
    on(Y,X),
    on_ground(X).
on_ground(X) :-
    \+ off_ground(X).
off_ground(X) :-
    on(X,Y).
on(c,b).
on(b,a).
```

go_well_together (X,Y) :-
\+ incompatible(X,Y).
incompatible(X,Y) :-
\+ likes(X,Y).
incompatible(X,Y) :-
\+ likes(Y,X).
likes(X,Y) :harmless(Y).
likes(X,Y) :eats (X,Y).
harmless(rabbit).
eats(python,rabbit).

# father (X,Y) :parent (X,Y), \+ mother (X,Y). 

parent (a,b). parent (c, b).
mother (a, b).
father (X,Y) :-
parent (X,Y),

$$
\backslash+\operatorname{mother}(\mathrm{X}, \mathrm{Y}) .
$$

mother (X,Y) :-
parent (X,Y),

$$
\text { \+ father }(X, Y)
$$

parent (a,b).
parent (c, b).

$$
\begin{aligned}
& \text { on_top }(X):- \\
& \quad \backslash+\text { blocked }(X) .
\end{aligned}
$$

$$
\begin{gathered}
\text { blocked }(X):- \\
\text { on }(Y, X) .
\end{gathered}
$$

on ( $a, b$ ).
\%---------------------
| ? - + on_top (b).
| ? $\quad$ + + on_top $(X)$.

# Logic and Grammars: Overview 

- Context free languages;
- Context sensitive languages;
- Definite Clause Grammars (DCGs);
- DCGs and Prolog.


## Context free Ianguages

- A context free grammar is a triple $\langle N, T, P\rangle$ where:
- $N$ is a finite set of non-terminals;
- $T$ is a finite set of terminals (and $N \cap T=\emptyset)$;
- $P \subseteq N \times(N \cup T)^{*}$ is a finite set of production rules.
- Examples of production rules:

$$
\begin{aligned}
\langle\text { expr }\rangle & \rightarrow\langle\text { expr }\rangle+\langle\text { expr }\rangle \\
\langle\text { sent }\rangle & \rightarrow\langle n p\rangle\langle v p\rangle
\end{aligned}
$$

## Derivations

- Let $\alpha, \beta, \gamma \in(N \cup T)^{*}$. We say that $\alpha A \gamma$ directly derives $\alpha \beta \gamma$ iff $A \rightarrow \beta \in P$. Denoted

$$
\alpha A \gamma \Rightarrow \alpha \beta \gamma
$$

- We say that $\alpha_{1}$ derives $\alpha_{n}$ iff there exists a sequence $\alpha_{1} \Rightarrow \alpha_{2}, \alpha_{2} \Rightarrow \alpha_{3}, \ldots, \alpha_{n-1} \Rightarrow \alpha_{n}$. Denoted

$$
\alpha_{1} \stackrel{*}{\Rightarrow} \alpha_{n}
$$

- A terminal string $\alpha \in T^{*}$ is in the language of $A$ iff $A \stackrel{*}{\Rightarrow} \alpha$.


## Example: Context free grammar

$\langle s e n t\rangle \rightarrow\langle n p\rangle\langle v p\rangle$
$\langle n p\rangle \rightarrow$ the $\langle n\rangle$
$\langle v p\rangle \rightarrow$ runs
$\langle n\rangle \rightarrow$ engine
$\langle n\rangle \rightarrow$ rabbit

## Naive implementation

$\operatorname{sent}(Z) \leftarrow \operatorname{append}(X, Y, Z), n p(X), v p(Y)$.
$n p([$ the $\mid X]) \leftarrow n(X)$.
$v p([$ runs $])$.
$n([$ engine $])$.
$n([r a b b i t])$.
$\operatorname{append}([], X s, X s)$.
append $([X \mid X s], Y s,[X \mid Z s]) \leftarrow$ $\operatorname{append}(X s, Y s, Z s)$.

## Usage of "Difference Lists"

- Assume that "-/2" denotes a partial function which given two strings $x_{1} \ldots x_{m-1} x_{m} \ldots x_{n}$ and $x_{m} \ldots x_{n}$ returns the string $x_{1} \ldots x_{m-1}$.
- Example

$$
\begin{aligned}
& \operatorname{sent}\left(X_{0}-X_{2}\right) \leftarrow n p\left(X_{0}-X_{1}\right), v p\left(X_{1}-X_{2}\right) \\
& \underbrace{x_{1} \ldots x_{i-1} \underbrace{x_{i} \ldots x_{j-1} \underbrace{x_{j} \ldots x_{k}}_{X_{2}}}_{X_{1}}}_{X_{0}}
\end{aligned}
$$

## Two Alternatives

$$
\begin{aligned}
& \text { sent }\left(X_{0}-X_{2}\right) \leftarrow n p\left(X_{0}-X_{1}\right), \text { vp }\left(X_{1}-X_{2}\right) . \\
& n p\left(X_{0}-X_{2}\right) \leftarrow C^{\prime}\left(X_{0}, \text { the }, X_{1}\right), n\left(X_{1}-X_{2}\right) . \\
& v p\left(X_{0}-X_{1}\right) \leftarrow '^{\prime}\left(X_{0}, \text { runs, } X_{1}\right) . \\
& n\left(X_{0}-X_{1}\right) \leftarrow C^{\prime}\left(X_{0}, \text { engine, } X_{1}\right) . \\
& n\left(X_{0}-X_{1}\right) \leftarrow C^{\prime}\left(X_{0}, \text { rabbits, } X_{1}\right) . \\
& C^{\prime}([X \mid Y], X, Y) .
\end{aligned}
$$

```
sent (X0-X ( 
np([the | X | ]- - X ) \leftarrown( 
vp([runs | X ] ]-X - ).
n([engine | }\mp@subsup{X}{1}{}]-\mp@subsup{X}{1}{})
n([rabbit | X | ]-X ( ).
```


## Partial deduction

```
grandparent(X,Y) :- parent(X,Z), parent(Z,Y).
```

parent (X,Y) :father (X,Y).
parent (X,Y) :mother (X,Y).
$\qquad$
grandparent(X,Y) :father (X,Z), parent(Z,Y).
grandparent(X,Y) :mother (X,Z), parent(Z,Y).
parent (X,Y) :father (X,Y).
parent (X,Y) :mother (X,Y).

## Context sensitive languages

- Some languages cannot be described by context free grammars. For instance

$$
\begin{aligned}
A B C & =\left\{a^{n} b^{n} c^{n} \mid n \geq 0\right\} \\
& =\{\epsilon, a b c, a a b b c c, a a a b b b c c c, \ldots\}
\end{aligned}
$$

- The language $A B C$ can be expressed in Prolog

$$
\begin{aligned}
& a b c\left(X_{0}-X_{3}\right) \leftarrow \\
& a\left(N, X_{0}-X_{1}\right), \\
& b\left(N, X_{1}-X_{2}\right), \\
& c\left(N, X_{2}-X_{3}\right) . \\
& a\left(0, X_{0}-X_{0}\right) . \\
& a\left(s(N),\left[a \mid X_{1}\right]-X_{2}\right) \leftarrow a\left(N, X_{1}-X_{2}\right) . \\
& b\left(0, X_{0}-X_{0}\right) . \\
& b\left(s(N),\left[b \mid X_{1}\right]-X_{2}\right) \leftarrow b\left(N, X_{1}-X_{2}\right) . \\
& c\left(0, X_{0}-X_{0}\right) . \\
& c\left(s(N),\left[c \mid X_{1}\right]-X_{2}\right) \leftarrow c\left(N, X_{1}-X_{2}\right) .
\end{aligned}
$$

## Definite Clause Grammars (DCGs)

- A Definite Clause Grammar is a triple $\langle N, T, P\rangle$ where
- $N$ is a finite/infinite set of atoms;
- $T$ is a finite/infinite set of terms (and $N \cap T=\emptyset)$;
- $P \subseteq N \times(N \cup T)^{*}$ is a finite set of production rules.


## Derivations

- Let $\alpha, \beta, \gamma \in(N \cup T)^{*}$. We say that $\alpha A \gamma$ directly derives $(\alpha \beta \gamma) \theta$ iff $A^{\prime} \rightarrow \beta \in P$ and $m g u\left(A, A^{\prime}\right)=\theta$. Denoted

$$
\alpha A \gamma \Rightarrow(\alpha \beta \gamma) \theta
$$

- We say that $\alpha_{1}$ derives $\alpha_{n}$ (denoted $\left.\alpha_{1} \stackrel{*}{\Rightarrow} \alpha_{n}\right)$ iff there exists a sequence

$$
\alpha_{1} \Rightarrow \alpha_{2}, \alpha_{2} \Rightarrow \alpha_{3}, \ldots, \alpha_{n-1} \Rightarrow \alpha_{n}
$$

- A terminal string $\alpha \in T^{*}$ is in the language of $A$ iff $A \stackrel{*}{\Rightarrow} \alpha$.


## Example of DCG

$\operatorname{sent}(\mathrm{s}(\mathrm{X}, \mathrm{Y}))$--> np(X, N)\ vp(Y, N).
np(john, singular(3)) --> [john].
np(they, plural(3)) --> [they].
vp(run, plural(X)) --> [run].
vp(runs, singular(3)) --> [runs].

## Semantical (context sensitive) constraints

The following DCG describes the language $\left\{a^{2 n} b^{2 n} c^{2 n} \mid n \geq 0\right\}$
$a b c \quad-->\quad a(N), b(N), c(N)$, even(N).
a(0) --> [].
$a(s(N))$--> [a], a(N).
even(0) --> [].
even(s(s(N))) --> even(N).

## Note

- The language of even $(X)$ contains only the string $\epsilon!!!$
- This may be emphasized by writing

$$
a b c-->a(N), b(N), c(N),\{\operatorname{even}(N)\} .
$$

- and by defining even/1 as a logic program

$$
\begin{aligned}
& \operatorname{even}(0) . \\
& \operatorname{even}(s(s(X))) \leftarrow \operatorname{even}(X) .
\end{aligned}
$$

## DCGs and Prolog

- Every production rule in a DCG can be compiled into a Prolog clause;
- The resulting Prolog program can be used as a (top-down) parser for the language (cf. "recursive descent");


## Compilation

- Assume that $X_{0}, \ldots, X_{m}$ are distinct variables that do not occur in

$$
p\left(t_{1}, \ldots, t_{n}\right) \rightarrow T_{1}, \ldots, T_{m}
$$

- The Prolog program will then contain a clause

$$
p\left(t_{1}, \ldots, t_{n}, X_{0}, X_{m}\right) \leftarrow T_{1}^{\prime}, \ldots, T_{m}^{\prime} .
$$

where each $T_{i}^{\prime},(1 \leq i \leq m)$, is of the form

$$
\begin{gathered}
q\left(t_{1}, \ldots, t_{n}, X_{i-1}, X_{i}\right) \text { if } T_{i}=q\left(t_{1}, \ldots, t_{n}\right) \\
C^{\prime}\left(X_{i-1}, t, X_{i}\right) \text { if } T_{i}=[t] \\
T, X_{i-1}=X_{i} \text { if } T_{i}=\{T\} \\
X_{i-1}=X_{i} \text { if } T_{i}=[]
\end{gathered}
$$

## Example

$$
\begin{aligned}
& \text { sent --> np, vp. } \\
& \text { np --> [the], n. } \\
& \text { vp --> [runs]. } \\
& \text { n --> [boy]. }
\end{aligned}
$$

\% Translates into...
sent(S0,S2) :- np(S0,S1), vp(S1,S2).
np(S0,S2) :- 'C'(S0,the,S1), n(S1,S2).
vp(S0,S1) :- 'C'(S0,runs,S1).
n(S0,S1) :- 'C'(S0,boy,S1).
'C' ([X|Xs],X,Xs).

## Summary

- Logic programming can be used to define
- (Regular languages);
- Context free languages;
- Context sensitive languages;
- (Recursively enumerable languages).
- Definite Clause Grammars (DCGs);
- Compilation of DCGs into Prolog.



## Examples

\% Membership in a ordered binary tree member (X, node(Left, X, Right)). member (X, node(Left, Y, Right)) :$\mathrm{X}<\mathrm{Y}$, member (X, Left).
member (X, node(Left, Y, Right)) :X > Y, member(X, Right).
\% Property of being a father father (X) :parent(X, Y), male(X).

## General

- Prolog constructs the SLD(NF)-tree by a depth-first search in combination with backtracking.
- By means of cut (!) the user can prohibit the Prolog engine from exploring certain branches in the tree.
- Cut (!) may only occur in the righthand sides of clauses and can be viewed as a regular (nullary) atom.


## Principles

- Two principal uses
- Prune infinite and failed branches (green cut);
- Prune refutations (red cut).
- Acceptable "red cut":
- Prune multiple occurrences of the same answer.


## The Golden Rule

First write a correct program without cuts. Then add cuts in approprate places to improve the efficiency.


# Constraint Iogic programming 

- Constraints
- Operations on constraints
- Constraint Logic Programming
- Language
- Operational semantics
- Examples


## Constraint

Given a set of variables, a constraint is a restriction on the possible values of the variables.

## Example

Variables: $X, Y$.

Constraint I: $X^{2}+Y^{2} \leq 4$

Constraint II: $Y \geq 2-2 \cdot X$

## Solution

The constraint $X^{2}+Y^{2} \leq 4$ has a set of solutions - variable assignments when the constraint is true, e.g:
$\{X \mapsto 2, Y \mapsto 0\}$
$\{X \mapsto 0, Y \mapsto 2\}$ $\{X \mapsto 1, Y \mapsto 1\}$

A mapping from variables to values is called a valuation. A valuation where the constraint is true is called a solution.

## Domain of a constraint

Whether a constraint has a solution or not depends on the values that the variables can take.

The constraint $X^{2}=2$ has a real solution, but not an integer or a rational solution.

The set of all possible values of the variables is called the domain of the constraint.

## Conjunctive constraints

The conjunction of the primitive constraints
$X^{2}+Y^{2} \leq 4$ and $Y \geq 2-2 \cdot X$ is a new (conjunctive) constraint:


Sets of primitive constraints represent conjunctive constraints.

## Properties of constraints

A constraint is said to be satisfiable iff it has at least one solution.

A constraint $C_{1}$ implies a constraint $C_{2}$ (written $C_{1} \models C_{2}$ ) iff every solution of $C_{1}$ is also a solution of $C_{2}$.

Two constraints are equivalent if they have the same set of solutions.

## Optimal solutions

A solution $\sigma$ of a set of constraints $S$ is maximal subject to an expression $E$ if $\sigma(E)$ is greater than $\sigma^{\prime}(E)$ for any solution $\sigma^{\prime}$ of $S$.

## Example

The solution $\{X \mapsto 1.6, Y \mapsto-1.2\}$ is a maximal solution of

$$
\begin{aligned}
X^{2}+Y^{2} & \leq 4 \\
Y & \geq 2-2 \cdot X
\end{aligned}
$$

subject to $-Y$.

# Constraint Logic Programming 

sorted([]).
sorted([x]).
sorted([Fst,Snd|Rst]) :-
Fst $=<$ Snd, sorted([Snd|Rst]).
:- sorted([X1, X2, X3]).

ARITHMETIC ERROR!!!

## Language

- Functors and predicate symbols divided into:
- Uninterpreted symbols (Herbrand terms/atoms);
- Interpreted symbols (constraints).
- Special solvers handle constraints;
- SLD(NF)-resolution is used for Herbrand atoms;


## Language (cont'd.)

- A clause is an expression

$$
A_{0} \leftarrow C_{1}, \ldots, C_{m}, A_{1}, \ldots, A_{n}
$$

where

- $A_{0}, \ldots, A_{n}$ are Herbrand atoms;
$-C_{1}, \ldots, C_{m}$ are constraints.
- A goal is an expression

$$
\leftarrow C_{1}, \ldots, C_{m}, A_{1}, \ldots, A_{n}
$$

# CLP(X): A Family of Languages 

CLP(R) Linear equations over reals

CLP(Q) Linear equations over rationals

CLP(B) Booleans
CLP(FD) Finite domains

## Example CLP(R)

mortgage(Loan, Years,AInt,Bal, APay) :\{ Years>0,

Years <= 1,
Bal=Loan*(1+Years*AInt)-APay \}.
mortgage(Loan, Years,AInt, Bal,APay) :-
\{ Years>1,
NewLoan = Loan*(1+AInt)-APay,
Years1 = Years-1 \},
mortgage(NewLoan, Years1,AInt,Bal,APay).
?- mortgage (120000, 10, 0.1, 0, AnnPay).
AnnPay=19529.4
?- mortgage(Loan, 10, 0.1, 0, 19529.4).
Loan=120000
?- mortgage(Loan, 10, 0.1,0, AnnPay).
Loan=6.14457*AnnPay

## Resolution with constraints

A state is a pair $(G ; S)$ where $G$ is a goal, and $S$ is a constraint store. Given a program $P$ a derivation is a sequence of states:

- $(\leftarrow A, B ; S) \Rightarrow\left(\leftarrow A=A^{\prime}, B^{\prime}, B ; S\right)$ if $A^{\prime} \leftarrow B^{\prime} \in P$
- $(\leftarrow C, G ; S) \Rightarrow(\leftarrow G ;\{C\} \cup S)$
- $(G ; S) \Rightarrow$ fail if $\operatorname{sat}(S)=$ false;
- $(G ; S) \Rightarrow\left(G ; S^{\prime}\right)$ if $S$ and $S^{\prime}$ are equivalent.
- $(G ;\{X=t\} \cup S) \Rightarrow(G ; S)\{X / t\}$


## Example: Arithmetic

```
:- res(ser(r(10),r(20)),X).
```

$\operatorname{res}(r(X), Y):-$ $\{X=Y\}$.
res(cell(X),Y) :-
$\{Y=0\}$.
res(ser $(\mathrm{X} 1, \mathrm{X} 2), \mathrm{R}):-$
$\{R=R 1+R 2\}, r e s(X 1, R 1), r e s(X 2, R 2)$.
res(par(X1,X2),R) :-
$\{1 / R=1 / R 1+1 / R 2\}, \operatorname{res}(X 1, R 1), r e s(X 2, R 2)$.

# Modeling with Boolean constraints 

Boolean operations

| + | Disjunktion | $*$ | Conjunction |
| ---: | ---: | ---: | :--- |
| $=<$ | Implikation | $=:=$ | Equivalence |
| $\#$ | Exclusive or | $\sim$ | Negation |

MOS transistors


$$
\begin{aligned}
& \operatorname{nmos}(S, G, D):-\operatorname{sat}(S * G=:=D * G) . \\
& \operatorname{pmos}(S, G, D):-\operatorname{sat}\left(S * \sim_{G}=:=D * \sim G\right) .
\end{aligned}
$$

## Design of XOR-gate



$$
\begin{array}{r}
\text { circuit }(X, Y, Z):- \\
\operatorname{pmos}(X, Y, Z), \\
\operatorname{pmos}(1, X, T), \\
\\
\operatorname{nmos}(T, X, 0), \\
\operatorname{nmos}(T, Y, Z) \\
\operatorname{nmos}(Y, T, Z) \\
\\
\operatorname{pmos}(Y, X, Z)
\end{array}
$$

## Verification of correctness

?- circuit( $\mathrm{X}, \mathrm{Y}, \mathrm{Z}$ ), $\operatorname{taut}(\mathrm{Z}=:=\mathrm{X} \# \mathrm{Y}, 1)$. yes

## CLP with Finite Domains

- Constraints and constraint problems
- Primitive constraints
- CLP(FD)
- Optimization
- Global constraints


## Example



- $A, B$ and $C$ live in different houses
- C lives left of B
- B has two neighbors


## Constraint problem

- A constraint problem consists of a finite set of problem variables,
- Each variable takes its value from a given domain
- Constraints are relations that restrict the values that can be assigned to the problem variables


## Mathematical reformulation



- $A, B, C \in\{1,2,3\}$
- $A \neq B, A \neq C$ and $B \neq C$
- $C<B$
- $(A<B<C)$ or $(C<B<A)$


## Example

Two problem variables $X$ and $Y$ with the integer domains 5..10 and 1..7. One constraint (relation) $\mathrm{X}<\mathrm{Y}$ :


New domains imposed by the constraint:
$X$ in $5 . .6$
$Y$ in $6 . .7$

## Operations on constraints

- Satisfiability: Does a given set of constraint have at least one solution?
- Entailment: Is every solution of a set $S$ of constraints also a solution of a constraint $C$ (denoted $S \models C$ )?
- Equality: Do two sets of constraints have the same set of solutions?
- Optimality: Find the best solution (given some criterion of optimality)
- Simplification: Given a set $S$ of constraints, find a simpler set of constraints $S^{\prime}$ equivalent to $S$.


## Primitive Finite Domain constraints

| ? $-X$ in 3..8.
$X$ in 3.. 8
| ? -X in 3..8, Y in 1..4, $\mathrm{Z} \#=\mathrm{X}+\mathrm{Y}$.
$X$ in 3..8,
Y in 1..4,
Z in 4..12
| ? - $X$ in 5..10, $Y$ in 1..7, $X$ \#< $Y$.
$X$ in 5..6,
$Y$ in 6..7

## Domains vs solutions

Note that domains are not identical to solutions:

?- $X$ in $5 . .10$, $Y$ in 1..7, $X$ \#< $Y$.

Produces the domains:

X in 5..6.
Y in 6..7.

But the domains contain all solutions:
$X=5, Y=6$
$X=5, Y=7$
$X=6, Y=7$

## More examples

$$
\begin{aligned}
& \text { | ?- } X \text { in 0..9, } Y \text { in 0..1, } X \text { \#< } Y \text {. } \\
& X=0 \text {, } \\
& Y=1 \\
& \text { | ?- } X \text { in 4..6, } Y \text { in 1..3, } X \text { \#< } Y \text {. } \\
& \text { no } \\
& \text { | ? - } X \text { in 1..12, } Y \text { in 1..12, } X \quad \#=2 * Y \text {. } \\
& \mathrm{X} \text { in 2..12, } \\
& Y \text { in } 1 . .6 \\
& \text { | ?- } X \text { in 1..2, } Y \text { in 1..2, } Z \text { in 1..2, } \\
& \mathrm{X} \# \backslash=\mathrm{Y}, \mathrm{X} \# \backslash=\mathrm{Z}, \mathrm{Y} \# \backslash=\mathrm{Z} . \\
& X \text { in 1..2, } \\
& Y \text { in 1..2, } \\
& \text { Z in 1..2 }
\end{aligned}
$$

Parallel declaration of domains
| ?- domain([X,Y,Z], 0, 9).

## Labeling

Domains approximate solutions...
| ?- $X$ in 1..2, $Y$ in 1..3, $X$ \#< $Y$.
X in 1..2,
$Y$ in 2..3

Systematically assign values to a variable from its domain.
| ?- X in 1..2, Y in 1..3, X \#< Y , labeling([],[X,Y]).
$\mathrm{X}=1, \mathrm{Y}=2$
$X=1, \quad Y=3$
$X=2, \quad Y=3$
| ?- X in $1 . .12$, Y in $1 . .12$, X \#= $2 * \mathrm{Y}$,
labeling ([],[X,Y]).
$\mathrm{X}=2, \quad \mathrm{Y}=1$
$X=4, Y=2$

## CLP (X)

A logic program is a set of rules

$$
A_{0}:-A_{1}, \ldots, A_{n}
$$

or facts

$$
A_{0}
$$

where $A_{0}, A_{1}, \ldots, A_{n}$ are atomic formulas; i.e. formulas of the form $p\left(t_{1}, \ldots, t_{n}\right)$.

Note: A constraint is an atomic formula!

A constraint logic program is a logic program where some of $A_{1}, \ldots, A_{n}$ may be (some pre-defined) constraints over some algebraic structure $X$.

## CLP(X)

- $\operatorname{CLP}(R)$, reals
- CLP(Q), rational numbers
- CLP(B), Boolean values
- CLP(FD), finite domains
- CLP(Sets), sets


## CLP(FD)

1. queens ( $\mathrm{N}, \mathrm{L}$ ) :-
2. length(L, N),
3. domain(L, 1, N),
4. safe(L),
5. labeling([], L).
6. safe([]).
7. safe([X|Xs]) :-
8. safe_between(X, Xs, 1),
9. safe(Xs).
10. safe_between(X, [], M).
11. safe_between(X, [Y|Ys], M) :-
12. no_attack(X, Y, M),
13. M1 is $\mathrm{M}+1$,
14. safe_between(X, Ys, M1).
15. no_attack(X, Y, N) :-
16. $\mathrm{X} \# \backslash=\mathrm{Y}, \mathrm{X}+\mathrm{N} \# \backslash=\mathrm{Y}, \mathrm{X}-\mathrm{N} \# \backslash=\mathrm{Y}$.

## General Strategy

1. solution(L) :-
2. create_variables(L),
3. constrain_variables(L),
4. solve_constraints(L).

## Optimization

$$
\begin{aligned}
\text { I ?- } & \mathrm{X} \text { in } 1 . .9, \mathrm{Y} \text { in } 4 \ldots 6, \mathrm{Z} \#=\mathrm{X}-\mathrm{Y}, \\
& \operatorname{labeling}([\operatorname{maximize}(\mathrm{Z})],[\mathrm{X}, \mathrm{Y}]) .
\end{aligned}
$$

1. items $(A, B, C, S, P)$ :-
2. domain ([A, B , C] , 0, 10) ,
3. $A S$ \# $=2 * \mathrm{~A}, \mathrm{AP} \#=3 * \mathrm{~A}$,
4. $B S$ \# $=3 * B, B P \#=4 * B$,
5. CS \#= $7 * C, C P$ \# $=10 * C$,
6. S \#>= AS+BS+CS,
7. $P$ \#= AP+BP+CP,
8. labeling([maximize (P)],[P, S, A, B , C]).

## Global Constraints

all_different $\left(\left[X_{1}, \ldots, X_{n}\right]\right)$

1. $\operatorname{smm}([S, E, N, D, M, O, R, Y]):-$
2. domain([S,E,N,D,M,O,R,Y], O, 9),
3. S \#> 0, M \#> 0,
4. all_different([S,E,N,D,M,O,R,Y]),
5. $\quad \operatorname{sum}(S, E, N, D, M, O, R, Y)$,
6. labeling([], [S,E,N,D,M,O,R,Y]).
7. sum(S, E, N, D, M, O, R, Y) :-
8. 

$1000 * S+100 * E+10 * N+D$
9.
$+1000 * \mathrm{M}+100 * \mathrm{O}+10 * \mathrm{R}+\mathrm{E}$
10. \#= $10000 * \mathrm{M}+1000 * 0+100 * \mathrm{~N}+10 * E+\mathrm{Y}$.
cumulative(Ss,Ds,Rs,L)
| ?- domain([S1,S2,S3],0,4), S1 \#< S3, cumulative([S1, S2, S3], [3, 4, 2], $[2,1,3], 3)$, labeling ([], [S1, S2, S3]).

Resource allocation

1. shower (S, Done) :-
2. 

$D=[5,3,8,2,7,3,9,3,3,5,7]$,
3. $R=[1,1,1,1,1,1,1,1,1,1,1]$,
4. length (D, N),
5. length (S, N),
6. domain(S, 0, 100),
7. Done in 0..100,
8. ready (S, D, Done),
9. cumulative (S, D, R, 3),
10. labeling([minimize(Done)], [Done|S]).
11. ready([], [], _).
12. ready([S|Ss], [D|Ds], Done) :-
13. Done \#>= S+D,
14. ready(Ss, Ds, Done).
element $\left(X,\left[X_{1}, \ldots, X_{n}\right], Y\right)$
| ?- element(X, [1,2,3,5], Y).
| ?- X in 2..3, element(X, [1, X, 4, 5], Y).
circuit $\left(\left[X_{1}, \ldots, X_{n}\right]\right)$

## Traveling Salesman

|  | $X_{1}$ | $X_{2}$ | $X_{3}$ | $X_{4}$ | $X_{5}$ | $X_{6}$ | $X_{7}$ |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| $X_{1}$ | - | 4 | 8 | 10 | 7 | 14 | 15 |
| $X_{2}$ | 4 | - | 7 | 7 | 10 | 12 | 5 |
| $X_{3}$ | 8 | 7 | - | 4 | 6 | 8 | 10 |
| $X_{4}$ | 10 | 7 | 4 | - | 2 | 5 | 8 |
| $X_{5}$ | 7 | 10 | 6 | 2 | - | 6 | 7 |
| $X_{6}$ | 14 | 12 | 8 | 5 | 6 | - | 5 |
| $X_{7}$ | 15 | 5 | 10 | 8 | 7 | 5 | - |

## Traveling Salesman (cont'd)

1. tsp(Cities, Cost) :-
2. Cities $=[\mathrm{X} 1, \mathrm{X} 2, \mathrm{X} 3, \mathrm{X} 4, \mathrm{X} 5, \mathrm{X} 6, \mathrm{X} 7]$,
3. element (X1, [ $0,4,8,10,7,14,15], C 1)$,
4. element (X2, [ 4, 0, 7, 7,10,12, 5], C2),
5. element (X3, [ 8, 7, 0, 4, 6, 8,10], C3),
6. element (X4, [10, 7, 4, 0, 2, 5, 8], C4),
7. element (X5, [ 7,10, 6, 2, 0, 6, 7], C5),
8. element (X6, [14, 12, 8, 5, 6, 0, 5], C6),
9. element (X7, [15, 5, 10, 8, 7, 5, 0], C7),
10. Cost \#= C1+C2+C3+C4+C5+C6+C7,
11. circuit(Cities),
12. labeling([minimize(Cost)], Cities).

## Deductive Databases: Overview

- Top-down evaluation;
- Relational databases;
- Bottom-up evaluation;
- "Magic templates"


# Logic programs as Databases 

- Powerful language for representation of relational data.
- Explicit data
- Views
- Queries
- Integrity constraints
- How to compute answers to database queries?
- Does not address issues such as concurrency control, updates, crashes etc.


## Top-down $\Rightarrow$ Recomputation

```
path(X,Y) :- edge(X,Y).
path(X,Z) :- edge(X,Y), path(Y,Z).
edge(a,b).
edge(b,c).
edge(a,c).
...
```


## Top-down $\Rightarrow$ Infinite computations

$\operatorname{path}(X, Y)$ :- edge(X,Y).
$\operatorname{path}(X, Z):-\operatorname{path}(X, Y), \operatorname{edge}(Y, Z)$.
edge ( $a, b$ ).
edge ( $b, a$ ).
edge (b, c).

# Properties: Top-down 

- Advantages:
- Efficient handling of search space;
- Goal-directed (Backward-chaining);
- Disadvantages:
- Termination;
- Recomputations;


## How to compute database queries?

Example:

| Father |  | Mother |  |
| :---: | :---: | :---: | :---: |
| X | Y | X | Y |
| tom | mary | mary | billy |
| john | tom | kate | tom |
| : | : | : | : |

New derived relations using relational algebra:

$$
\begin{aligned}
P & :=F(X, Y) \cup M(X, Y) \\
G P & :=\pi_{X, Z}(P(X, Y) \bowtie P(Y, Z))
\end{aligned}
$$

## Bottom-up evaluation (Cf. $T_{P}$ )

$$
\begin{aligned}
S_{P}(X)= & \\
\left\{A_{0} \theta \mid\right. & A_{0} \leftarrow A_{1}, \ldots, A_{n} \in P \text { and } \\
& A_{1}^{\prime}, \ldots, A_{n}^{\prime} \in X \text { and } \\
& \left.m g u\left\{A_{1}=A_{1}^{\prime}, \ldots, A_{n}=A_{n}^{\prime}\right\}=\theta\right\}
\end{aligned}
$$

Naive evaluation
fun naive( $P$ )
begin
$x:=$ facts $(P)$;
repeat

$$
\begin{aligned}
& y:=x ; \\
& x:=S_{P}(y) ;
\end{aligned}
$$

until $x=y$;
return $x$;
end

## Bottom-up evaluation (cont'd.)

$\Delta S_{P}(X, \Delta X)=$

$$
\begin{aligned}
\left\{A_{0} \theta \mid\right. & A_{0} \leftarrow A_{1}, \ldots, A_{n} \in P \text { and } \\
& A_{1}^{\prime}, \ldots, A_{n}^{\prime} \in X, \exists A_{i}^{\prime} \in \triangle X \text { and } \\
& \left.m g u\left\{A_{1}=A_{1}^{\prime}, \ldots, A_{n}=A_{n}^{\prime}\right\}=\theta\right\}
\end{aligned}
$$

## Semi-naive evaluation

fun seminaive ( $P$ )
begin
$\Delta x:=\operatorname{facts}(P) ;$
$x:=\Delta x$;
repeat
$\Delta x:=\Delta S_{P}(x, \Delta x) \backslash x ;$
$x:=x \cup \Delta x ;$
until $\Delta x=\emptyset$;
return $x$;
end

# Properties: Bottom-up 

- Advantages:
- Termination;
- Re-use of already computed results;
- Disadvantages:
- Not goal-directed;
- Termination;


## Magic Templates

Let $\operatorname{magic}(P)$ be the least program such that if $A_{0} \leftarrow A_{1}, \ldots, A_{n} \in P$ then:

- $A_{0} \leftarrow \operatorname{call}\left(A_{0}\right), A_{1}, \ldots, A_{n} \in \operatorname{magic}(P)$
- $\operatorname{call}\left(A_{i}\right) \leftarrow \operatorname{call}\left(A_{0}\right), A_{1}, \ldots, A_{i-1} \in$ $\operatorname{magic}(P)$

In addition $\operatorname{call}(A) \in \operatorname{magic}(P)$ if $\leftarrow A$.

Compute naive(magic $(P)$ ).

## Example

```
%-----------ORIGINAL PROGRAM----------------
p(X,Y) :- e(X,Y).
p(X,Z) :- p(X,Y), e(Y,Z).
e(a,b).
e(b,a).
e(b,c).
    :- p(a,X).
%--------------MAGIC PROGRAM---------------
p(X,Y) :- call(p(X,Y)), e(X,Y).
p(X,Z) :- call(p(X,Z)), p(X,Y), e(Y,Z).
e(a,b) :- call(e(a,b)).
e(b,a) :- call(e(b,a)).
e(b,c) :- call(e(b,c)).
%
call(e(X,Y)) :- call(p(X,Y)).
call(p(X,Y)) :- call(p(X,Z)).
call(e(Y,Z)) :- call(p(X,Z)), p(X,Y).
%
call(p(a,X)).
```


# Bottom-up with Magic Templates 

- Advantages:
- Termination;
- Re-use of results;
- Goal-directed;
- Disadvantages:
- Sometimes slower than Prolog (when Prolog terminates);


# Logic programming with Equations 

- What is equality?
- E-unification.
- Logic programs with Equations
- SLDE-resolution


## What is equality?

We sometimes want to express that two terms should be interpreted as the same object.

## Example

Let $\Gamma$ be:

$$
\begin{aligned}
& \text { person }(X) \leftarrow \text { female }(X) . \\
& \text { female(queen). } \\
& \text { silvia } \doteq \text { queen } .
\end{aligned}
$$

Then $\Gamma \vDash$ person(silvia).

## Equations

An equation is an atom $s \doteq t$ where $s$ and $t$ are terms.

The predicate $\doteq$ is always interpreted as the identity relation.

That is, $\Im \models_{\sigma} s \doteq t$ iff $\sigma_{\Im}(s)=\sigma_{\Im}(t)$.

Example

$$
\begin{aligned}
X+0 & \doteq X . \\
X+s(Y) & \doteq \\
1 & \doteq s(X+Y) . \\
2 & \doteq 1+1 . \\
3 & \doteq \\
\doteq & 10) .
\end{aligned}
$$

## Equality theory

$E \vdash s \doteq t:$ " $s \doteq t$ is derived from $E$ "

$$
\begin{gathered}
\{\ldots, s \doteq t, \ldots\} \vdash s \doteq t \\
E \vdash s \doteq s \\
\frac{E \vdash s \doteq t}{E \vdash s \sigma \doteq t \sigma} \\
\frac{E \vdash s \doteq t}{E \vdash t \doteq s} \\
\frac{E \vdash r \doteq s \quad E \vdash s \doteq t}{E \vdash r \doteq t} \\
\frac{E \vdash s_{1} \doteq t_{1} \cdots E \vdash s_{n} \doteq t_{n}}{E \vdash f\left(s_{1}, \ldots, s_{n}\right) \doteq f\left(t_{1}, \ldots, t_{n}\right)} \\
* * * \\
s \equiv_{E} t \text { ff } E \vdash s \doteq t
\end{gathered}
$$

## Theorem

The relation $\equiv_{E}$ is an equality relation.

Theorem
$E \vDash s \doteq t$ iff $s \equiv_{E} t($ iff $E \vdash s \doteq t)$.

E-unification

Two terms $s$ and $t$ are $E$-unifiable iff $s \theta \equiv_{E} t \theta$. The substitution $\theta$ is called an $E$-unifier.

## Problem

- E-unification is undecidable;
- In general there is no single "most general unifier" but only "complete sets of E-unifiers";
- This set may be infinite.


## Unification. . .

...can be carried out using e.g. narrowing.

# Logic programs with Equations 

 Programs consist of two components- A set of definite clauses that do not include the predicate symbol $\doteq / 2$;
- A set of equations;


## Observation

Herbrand interpretations are uninteresting!

## Patch

Consider interpretations whose domain consists of sets (equivalence classes) of ground terms.

Every equivalence class consists of "equivalent term".

Interpretations with domain $U_{P} / \equiv_{E}$ are of special interest.

Let $\Im$ be an interpretation where $|\Im|=U_{P} / \equiv_{E}$ :
That is, $\bar{s}=\left\{t \in U_{P} \mid E \vdash s \doteq t\right\}$.

## Theorem

$$
\begin{array}{lll}
\Im \models s \doteq t & \text { iff } & \bar{s}=\bar{t} \\
& \text { iff } & s \equiv t \\
& \text { iff } & E \models s \doteq t
\end{array}
$$

NB: Herbrand interpretations as a special case!

## The Least Model

Every program $P, E$ has a least model $M_{P, E}$ :

$$
P, E \models p\left(t_{1}, \ldots, t_{n}\right) \text { iff } \overline{p\left(t_{1}, \ldots, t_{n}\right)} \in M_{P, E}
$$

## Fixed point semantics

$$
\begin{aligned}
T_{P, E}(x):=\left\{\begin{array}{lll}
\bar{A} & \mid & A \leftarrow B_{1}, \ldots, B_{n} \in \operatorname{ground}(P) \\
& \wedge \overline{B_{1}}, \ldots, \overline{B_{n}} \in x
\end{array}\right)
\end{aligned}
$$

## SLDE-Resolution

Given a goal

$$
\leftarrow A_{1}, \ldots, A_{i-1}, A_{i}, A_{i+1}, \ldots, A_{n}
$$

with selected literal $A_{i}$. If

- $H \leftarrow B_{1}, \ldots, B_{m}$ is a renamed program clause
- $H$ and $A_{i}$ have a non-empty set $\Theta$ of $E$-unifiers
- $\theta \in \Theta$
then

$$
\leftarrow\left(A_{1}, \ldots, A_{i-1}, B_{1}, \ldots, B_{m}, A_{i+1}, \ldots, A_{n}\right) \theta
$$

is a new goal.

## Theorem [Soundness]

If $\leftarrow A_{1}, \ldots, A_{n}$ has a computed answer substitution $\theta$ then $P, E \models \forall\left(A_{1} \wedge \cdots \wedge A_{n}\right) \theta$.

## Theorem [Completeness]

Similar to SLD-resolution.

