

Constraint Logic Programming

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INRIA – Projet CONTRAINTES

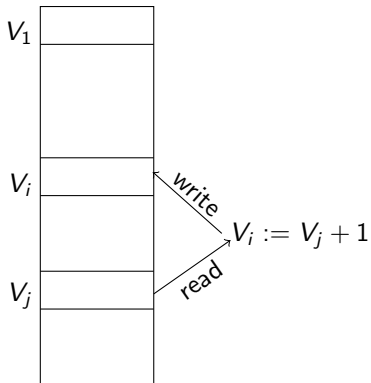
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Part I: CLP - Introduction and Logical Background

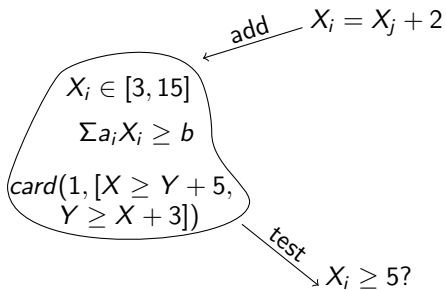
- 1 The Constraint Programming paradigm
- 2 Examples and Applications
- 3 First Order Logic
- 4 Models
- 5 Logical Theories

The Constraint programming Machine

memory of values
programming variables



memory of constraints
mathematical variables



The Paradigm of Constraint Programming

Program = Logical Formula

Axiomatization:
"Domain of discourse" \mathcal{X} ,
Model of the problem P

Execution = Proof search

Constraint satisfiability,
Logical resolution principle

Class of languages $\text{CLP}(\mathcal{X})$ parametrized by \mathcal{X} :

- Primitive Constraints over \mathcal{X}

$$U = R * I$$

- Relations defined by logical formulas

$$\forall x, y \text{ path}(x, y) \Leftrightarrow \text{edge}(x, y) \vee \exists z (\text{edge}(x, z) \wedge \text{path}(z, y))$$

Languages for defining new relations

- First-order **logic** predicate calculus

$$\forall x, y \text{ path}(x, y) \Leftrightarrow \text{edge}(x, y) \vee \exists z(\text{edge}(x, z) \wedge \text{path}(z, y))$$

- Prolog/**CLP(\mathcal{X})** clauses

$\text{path}(X, Y) :- \text{edge}(X, Y) .$

$\text{path}(X, Y) :- \text{edge}(X, Z) , \text{path}(Z, Y) .$

- Concurrent constraint process languages **CC(\mathcal{X})**

Process $A = c \mid p(x) \mid (A \parallel A) \mid A + A \mid \text{ask}(c) \rightarrow A \mid \exists x A$
 $\text{path}(X, Y) :: \text{edge}(X, Y) + \exists Z(\text{edge}(X, Z) \parallel \text{path}(Z, Y))$

- Constraint **libraries** in object-oriented/functional/imperative languages (ILOG, Koalog, etc.).

CLP(FD) N-Queens Problem

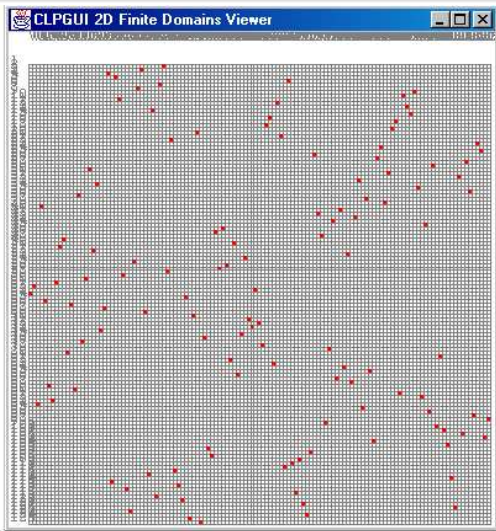
GNU-Prolog program:

```
queens(N,L):-  
    length(L,N),  
    fd_domain(L,1,N),  
    safe(L),  
    fd_labeling(L,first_fail).  
safe([]).  
safe([X|L]):-  
    noattack(L,X,1),  
    safe(L).  
noattack([],_,_).  
noattack([Y|L],X,I):-  
    X#\=Y,  
    X#\=Y+I,  
    X+I#\=Y,  
    I1 is I+1,  
    noattack(L,X,I1).
```

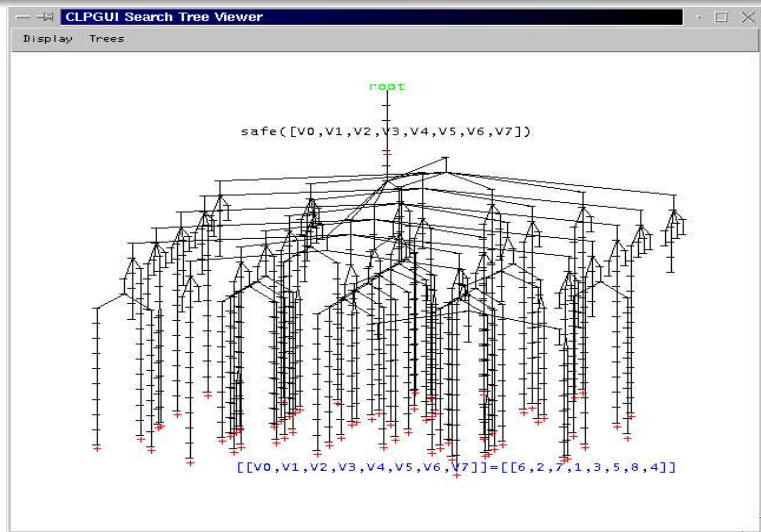
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safe([X|L]):-  
    noattack(L,X,1),  
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noattack([],_,_).  
noattack([Y|L],X,I):-  
    X#\=Y,  
    X#\=Y+I,  
    X+I#\=Y,  
    I1 is I+1,  
    noattack(L,X,I1).
```



Search space of all solutions



CLPGUI Demo

Successes in combinatorial search problems

Job shop scheduling, resource allocation, graph coloring,...

- **Decision Problems:** existence of a solution (of given cost)
in **P**: if algorithm of polynomial time complexity
in **NP**: if *non-deterministic* algorithm of polynomial complexity.
NP-complete if polynomial encoding of any other NP problem
- **Optimisation Problems:** computation of a solution of optimal cost
NP-hard if the decision problem is NP-complete
- The size of the search space does not tell the complexity of the problem
Sorting n elements in $O(n \log n)$, search space in $n!$...
SAT over n Boolean in $O(2^n)$, search space in 2^n .

Workplan of the Lecture

- 1 Introduction to CLP, operational semantics, examples
- 2 CLP - Fixpoint and logical semantics
- 3 CSP resolution - simplification and domain reduction
- 4 Symmetries - variables, values, breaking
- 5 Global constraints and graph properties
- 6 CC - Examples, operational and denotational semantics
- 7 CC - Linear Logic semantics
- 8 LCC, CHR, SiLCC

Written exam + **Programming project**

Hot Research Topics in Constraint Programming

- Combinatorial Benchmarks (open shop 6x6, social golfer,...)
Global constraints
Search procedures, randomization
Hybridization of algorithms CP, MILP, local search
Symmetry detection and breaking
- Easily extensible CP languages
Adaptive solving strategies
Automatic synthesis of constraint solvers
- New applications in Bioinformatics

⇒ Internships

First-Order Terms

Alphabet:

set of variables V ,

set of constant and function symbols S_F , given with their arity α

The set T of *first-order terms* is the least set satisfying

- 1 $V \subset T$
- 2 if $f \in S_F$, $\alpha(f) = n$, $M_1, \dots, M_n \in T$
then $f(M_1, \dots, M_n) \in T$

First-order Formulas

Alphabet: set S_P of **predicate symbols**.

Atomic propositions: $p(M_1, \dots, M_n)$ where $p \in S_P$, $M_1, \dots, M_n \in T$.

Formulas: $\neg\phi$, $\phi \vee \psi$, $\exists x \phi$

The other logical symbols are defined as abbreviations:

$$\phi \Rightarrow \psi = \neg\phi \vee \psi$$

$$\text{true} = \phi \Rightarrow \phi$$

$$\text{false} = \neg\text{true}$$

$$\phi \wedge \psi = \neg(\phi \Rightarrow \neg\psi)$$

$$\phi \equiv \psi = (\phi \Rightarrow \psi) \wedge (\psi \Rightarrow \phi)$$

$$\forall x \phi = \neg \exists x \neg \phi$$

Clauses

A *literal* L is either an atomic proposition, A , (called a *positive literal*), or the negation of an atomic proposition, $\neg A$ (called a *negative literal*).

A *clause* is a disjunction of universally quantified literals,

$$\forall(L_1 \vee \dots \vee L_n),$$

A *Horn clause* is a clause having at most one positive literal.

$$\neg A_1 \vee \dots \vee \neg A_n$$

$$A \vee \neg A_1 \vee \dots \vee \neg A_n$$

Interpretations

An interpretation $\langle D, [] \rangle$ is a mathematical structure given with

- a domain D ,
- distinguished elements $[c] \in D$ for each constant $c \in S_F$,
- operators $[f] : D^n \rightarrow D$ for each function symbol $f \in S_F$ of arity n .
- relations $[p] : D^n \rightarrow \{\text{true}, \text{false}\}$ for each predicate symbol $p \in S_P$ of arity n

Valuation

A **valuation** is a function $\rho : V \rightarrow D$ extended to terms by morphism

- $[x]_\rho = \rho(x)$ if $x \in V$,
- $[f(M_1, \dots, M_n)]_\rho = [f]([M_1]_\rho, \dots, [M_n]_\rho)$ if $f \in S_F$

The **truth value of an atom** $p(M_1, \dots, M_n)$ in an interpretation $I = \langle D, [] \rangle$ and a valuation ρ is the boolean value $[p]([M_1]_\rho, \dots, [M_n]_\rho)$.

The **truth value of a formula** in I and ρ is determined by truth tables and

$[\exists x \phi]_\rho = \text{true}$ if $[\phi[d/x]]_\rho = \text{true}$ for some $d \in D$, false otherwise.

$[\forall x \phi]_\rho = \text{true}$ if $[\phi[d/x]]_\rho = \text{true}$ for every $d \in D$, false otherwise.

Models

- An interpretation I is a *model* of a closed formula ϕ , $I \models \phi$, if ϕ is true in I .
- A closed formula ϕ' is a *logical consequence* of ϕ closed, $\phi \models \phi'$, if every model of ϕ is a model of ϕ' .
- A formula ϕ is *satisfiable in an interpretation I* if $I \models \exists(\phi)$, (e.g. $\mathcal{Z} \models \exists x x < 0$)
 ϕ is *valid in I* if $I \models \forall(\phi)$.
- A formula ϕ is *satisfiable* if $\exists(\phi)$ has a model (e.g. $x < 0$)
- A formula is *valid*, noted $\models \phi$, if every interpretation is a model of $\forall(\phi)$ (e.g. $p(x) \Rightarrow \exists y p(y)$)

Proposition

For closed formulas, $\phi \models \phi'$ iff $\models \phi \Rightarrow \phi'$.

Herbrand's Domain \mathcal{H}

Domain of closed terms $T(S_F)$ "Syntactic" interpretation

$$[c] = c$$

$$[f(M_1, \dots, M_n)] = f([M_1], \dots, [M_n])$$

Herbrand's base $B_{\mathcal{H}} = \{p(M_1, \dots, M_n) \mid p \in S_P, M_i \in T(S_F)\}$

A *Herbrand's interpretation* is identified to a subset of $B_{\mathcal{H}}$
(the subset defines the atomic propositions which are true).

Herbrand's Models

Proposition

Let S be a set of clauses. S is *satisfiable* if and only if S has a *Herbrand's model*.

Proof.

Suppose I is a model of S : for every I -valuation ρ , for every clause $C \in S$, there exists a positive literal A (resp. negative literal $\neg A$) in C such that $I \models A\rho$ (resp. $I \not\models A\rho$).

Let I' be the Herbrand's interpretation defined by

$$I' = \{p(M_1, \dots, M_n) \in B_H \mid I \models p(M_1, \dots, M_n)\}.$$

For every Herbrand's valuation ρ' , there exists an I -valuation ρ such that $I \models A\rho$ iff $I' \models A\rho'$. Hence, for every clause, there exists a literal A (resp. $\neg A$) such that $I' \models A\rho'$ (resp. $I' \not\models A\rho'$).

Therefore I' is a Herbrand's model of S . □

Skolemization

- Put ϕ in prenex form (all quantifiers in the head)
- Replace an existential variable x by a term $f(x_1, \dots, x_k)$ where f is a new function symbol and the x_i 's are the universal variables before x

E.g. $\phi = \forall x \exists y \forall z p(x, y, z)$, $\phi^s = \forall x \forall z p(x, f(x), z)$.

Proposition

Any formula ϕ is satisfiable iff its Skolem's normal form ϕ^s is satisfiable.

If $I \models \phi$ then one can choose an interpretation of the Skolem's function symbols in ϕ^s according to the I -valuation of the existential variables of ϕ such that $I \models \phi^s$.

Conversely, if $I \models \phi^s$, the interpretation of the Skolem's functions in ϕ^s gives a valuation of the existential variables in ϕ s.t. $I \models \phi$.

Logical Theories

A *theory* is a formal system formed with

- logical axioms and inference rules

$$\neg A \vee A \text{ (excluded middle)} \quad A[x \leftarrow B] \Rightarrow \exists x A \text{ (substitution)}$$

$$\frac{A}{B \vee A} \text{ (Weakening)}$$

$$\frac{A \vee A}{A} \text{ (Contraction)}$$

$$\frac{B \vee A}{A \vee (B \vee C)} \text{ (Associativity)}$$

$$\frac{A \vee B \quad \neg A \vee C}{B \vee C} \text{ (Cut)}$$

$$\frac{A \Rightarrow B \quad x \notin V(B)}{\exists x A \Rightarrow B} \text{ (Existential introduction)}$$

- a set \mathcal{T} of non-logical axioms

Deduction relation: $\mathcal{T} \vdash \phi$ if the closed formula ϕ can be derived in \mathcal{T}

\mathcal{T} is *contradictory* if $\mathcal{T} \vdash \text{false}$, otherwise \mathcal{T} is *consistent*.

Validity

Theorem (Deduction theorem)

$\mathcal{T} \vdash \phi \Rightarrow \psi$ iff $\mathcal{T} \cup \{\phi\} \vdash \psi$.

The implication is immediate with the cut rule.

Conversely the proof is by induction on the derivation of the formula ψ . □

Theorem (Validity)

If $\mathcal{T} \vdash \phi$ then $\mathcal{T} \models \phi$.

By induction on the length of the deduction of ϕ . □

Corollary

If \mathcal{T} has a model then \mathcal{T} is consistent

We show the contrapositive: if \mathcal{T} is contradictory, then $\mathcal{T} \vdash \text{false}$, hence $\mathcal{T} \models \text{false}$, hence \mathcal{T} has no model.

Gödel's Completeness Theorem

Theorem

A theory is consistent iff it has a model.

The idea is to construct a Herbrand's model of the theory supposed to be consistent, by interpreting by true the closed atoms which are theorems of \mathcal{T} , and by false the closed atoms whose negation is a theorem of \mathcal{T} . For this it is necessary to extend the alphabet to denote domain elements by Herbrand terms. □

Corollary

$\mathcal{T} \models \phi$ iff $\mathcal{T} \vdash \phi$.

If $\mathcal{T} \models \phi$ then $\mathcal{T} \cup \{\neg\phi\}$ has no model, hence $\mathcal{T} \cup \{\neg\phi\} \vdash \text{false}$, and by the deduction theorem $\mathcal{T} \vdash \neg\neg\phi$, now by the cut rule with the axiom of excluded middle (plus weakening and contraction) we get $\mathcal{T} \vdash \phi$.

Axiomatic and Complete Theories

A theory \mathcal{T} is *axiomatic* if the set of non logical axioms is recursive (i.e. membership to this set can be decided by an algorithm).

Proposition

In an axiomatic theory \mathcal{T} , valid formulas, $\mathcal{T} \models \phi$, are recursively enumerable.

(expresses the feasibility of the **Logic Programming paradigm...**)

A theory is *complete* if for every closed formula ϕ , either $\mathcal{T} \vdash \phi$ or $\mathcal{T} \vdash \neg\phi$.

In a **complete axiomatic theory**, we can decide whether an arbitrary formula is satisfiable or not (**Constraint Satisfaction paradigm...**).

Compactness theorem

Theorem

$\mathcal{T} \models \phi$ iff $\mathcal{T}' \models \phi$ for some finite part \mathcal{T}' of \mathcal{T} .

By Gödel's completeness theorem, $\mathcal{T} \models \phi$ iff $\mathcal{T} \vdash \phi$.

As the proofs are finite, they use only a finite part of non logical axioms \mathcal{T} .

Therefore $\mathcal{T} \models \phi$ iff $\mathcal{T}' \models \phi$ for some finite part \mathcal{T}' of \mathcal{T} . □

Corollary

\mathcal{T} is consistent iff every finite part of \mathcal{T} is consistent.

\mathcal{T} is inconsistent iff $\mathcal{T} \vdash \text{false}$,

iff for some finite part \mathcal{T}' of \mathcal{T} , $\mathcal{T}' \vdash \text{false}$,

iff some finite part of \mathcal{T} is inconsistent. □

Coloring infinite maps with four colors

Let \mathcal{T} express the **coloriability with four colors** of an infinite planar graph G :

- $\forall x \bigvee_{i=1}^4 c_i(x)$,
- $\forall x \bigwedge_{1 \leq i < j \leq 4} \neg(c_i(x) \wedge c_j(x))$,
- $\bigwedge_{i=1}^4 \neg(c_i(a) \wedge c_i(b))$ for every adjacent vertices a, b in G .

Let \mathcal{T}' be any finite part of \mathcal{T} , and G' be the (finite) subgraph of G containing the vertices which appear in \mathcal{T}' . As G' is finite and planar it can be colored with 4 colors [Appel and Haken 76], thus \mathcal{T}' has a model.

Now as every finite part \mathcal{T}' of \mathcal{T} is satisfiable, we deduce from the compactness theorem that **\mathcal{T} is satisfiable**. Therefore every infinite planar graph can be colored with four colors.

Complete theory: Presburger's arithmetic

Complete axiomatic theory of $(\mathbb{N}, 0, s, +, =)$,

$$E_1 : \forall x \ x = x,$$

$$E_2 : \forall x \forall y \ x = y \rightarrow s(x) = s(y),$$

$$E_3 : \forall x \forall y \forall z \forall v \ x = y \wedge z = v \rightarrow (x = z \rightarrow y = v),$$

$$E_4, \Pi_1 : \forall x \forall y \ s(x) = s(y) \rightarrow x = y,$$

$$E_5, \Pi_2 : \forall x \ 0 \neq s(x),$$

$$\Pi_3 : \forall x \ x + 0 = x,$$

$$\Pi_4 : \forall x \ x + s(y) = s(x + y),$$

$$\Pi_5 : \phi[x \leftarrow 0] \wedge (\forall x \ \phi \rightarrow \phi[x \leftarrow s(x)]) \rightarrow \forall x \phi \text{ for every formula } \phi.$$

Note that $E_6 : \forall x \ x \neq s(x)$ and $E_7 : \forall x \ x = 0 \vee \exists y \ x = s(y)$ are provable by induction.

Gödel's Incompleteness Theorem

Peano's arithmetic contains moreover two axioms for \times :

$$\Pi_6: \forall x \ x \times 0 = 0,$$

$$\Pi_7: \forall x \forall y \ x \times s(y) = x \times y + x,$$

Theorem

Any consistent axiomatic extension of Peano's arithmetic is incomplete.

The idea of the proof, following the liar paradox of Epimenides (600 bc) which says: "I lie", is to construct in the language of Peano's arithmetic Π a formula ϕ which is true in the structure of natural numbers \mathbb{N} if and only if ϕ is not provable in Π . As \mathbb{N} is a model of Π , ϕ is necessarily true in \mathbb{N} and not provable in Π , hence Π is incomplete. \square

Corollary

*The structure $(\mathbb{N}, 0, 1, +, *)$ is not axiomatizable.*

Part II: Constraint Logic Programs

- 6 Constraint Languages
 - Decidability in Complete Theories
- 7 CLP(\mathcal{X})
 - Definition
 - Operational Semantics
- 8 CLP(\mathcal{H})
 - Prolog
 - Examples
- 9 CLP($\mathcal{R}, \mathcal{FD}, \mathcal{B}$)
 - CLP(\mathcal{R})
 - CLP(\mathcal{FD})
 - CLP(\mathcal{B})

Constraint Languages

Alphabet: set V of variables,
set S_F of constant and function symbols,
set S_C of predicate symbols containing true and $=$.

We assume a set of *basic constraints*, supposed to be closed by variable renaming, and to contain all atomic constraints.

The *language of constraints* is the closure by conjunction and existential quantification of the set of basic constraints.
Constraints will be denoted by c, d, \dots

Fixed Interpretation \mathcal{X}

Structure \mathcal{X} for interpreting the constraint language.

We assume that the constraint satisfiability problem, $\mathcal{X} \models^? \exists(c)$, is **decidable**.

This is equivalent to assume that \mathcal{X} is presented by an axiomatic theory \mathcal{T} satisfying:

- ① (soundness) $\mathcal{X} \models \mathcal{T}$
- ② (completeness for constraint satisfaction) for every constraint c , either $\mathcal{T} \vdash \exists(c)$, or $\mathcal{T} \vdash \neg\exists(c)$.

Clark's Equality Theory for the Herbrand domain

$$E_1 \quad \forall x \ x = x,$$

$$E_2 \quad \forall (x_1 = y_1 \wedge \dots \wedge x_n = y_n \rightarrow f(x_1, \dots, x_n) = f(y_1, \dots, y_n)),$$

$$E_3 \quad \forall (x_1 = y_1 \wedge \dots \wedge x_n = y_n \rightarrow p(x_1, \dots, x_n) \rightarrow p(y_1, \dots, y_n)),$$

$$E_4 \quad \forall (f(x_1, \dots, x_n) = f(y_1, \dots, y_n) \rightarrow x_1 = y_1 \wedge \dots \wedge x_n = y_n),$$

$$E_5 \quad \forall (f(x_1, \dots, x_m) \neq g(y_1, \dots, y_n)) \text{ for different function symbols } f, g \in S_F \text{ with arity } m \text{ and } n \text{ respectively,}$$

$$E_6 \quad \forall x \ M[x] \neq x \text{ for every term } M \text{ strictly containing } x.$$

Proposition

$\mathcal{H} \models CET.$

Proposition

Furthermore if the set of function symbols is infinite, CET is a *complete* theory.

CLP(\mathcal{X}) Programs

Alphabet V, S_F, S_C of constraint symbols.

Structure \mathcal{X} presented by a satisfaction complete theory \mathcal{T}

Alphabet S_P of *program predicate* symbols

A **CLP(\mathcal{X}) program** is a finite set of program clauses.

Program clause $\forall(A \vee \neg c_1 \vee \dots \vee \neg c_m \vee \neg A_1 \vee \dots \vee \neg A_n)$

$$A \leftarrow c_1, \dots, c_m \mid A_1, \dots, A_n$$

Goal clause $\forall(\neg c_1 \vee \dots \vee \neg c_m \vee \neg A_1 \vee \dots \vee \neg A_n)$

$$c_1, \dots, c_m \mid A_1, \dots, A_n$$

Operational semantics: CSLD Resolution

$$\frac{(p(t_1, t_2) \leftarrow c' | A_1, \dots, A_n)\theta \in P \quad \mathcal{X} \models \exists(c \wedge s_1 = t_1 \wedge s_2 = t_2 \wedge c')}{(c | \alpha, p(s_1, s_2), \alpha') \longrightarrow (c, s_1 = t_1, s_2 = t_2, c' | \alpha, A_1, \dots, A_n, \alpha')}$$

where θ is a renaming substitution of the program clause with new variables.

A **successful derivation** is a derivation of the form

$$G \longrightarrow G_1 \longrightarrow G_2 \longrightarrow \dots \longrightarrow c | \square$$

c is called a **computed answer constraint** for G .

Prolog as CLP(\mathcal{H})

The programming language *Prolog* is an implementation of CLP(\mathcal{H}) in which:

- the constraints are only equalities between terms,
- the selection strategy consists in solving the atoms from left to right according to their order in the goal,
- the search strategy consists in searching the derivation tree *depth-first* by *backtracking*.

Only constants: Deductive Databases

```
gdfather(X,Y):-father(X,Z),parent(Z,Y).  
gdmother(X,Y):-mother(X,Z),parent(Z,Y).  
parent(X,Y):-father(X,Y).  
parent(X,Y):-mother(X,Y).  
father(alphonse,chantal).  
mother(emilie,chantal).  
mother(chantal,julien).  
father(julien,simon).
```

```
| ?- gdfather(X,Y).  
X = alphonse, Y = julien ? ;  
no
```

```
| ?- gdmother(X,Y).  
X = emilie, Y = julien ? ;  
X = chantal, Y = simon ? ;  
no
```

Lists

```

member(X, cons(X, L)).
member(X, cons(Y, L)) :- member(X, L).

| ?- member(X, cons(a, cons(b, cons(c, nil)))).
X = a ? ;
X = b ? ;
X = c ? ;
no
| ?- member(X, Y).
Y = cons(X, _A) ? ;
Y = cons(_B, cons(X, _A)) ? ;
Y = cons(_C, cons(_B, cons(X, _A))) ?
yes

```

Appending lists

```
append([],L,L).
append([X|L],L2,[X|L3]):-append(L,L2,L3).
```

```
| ?- append([a,b],[c,d],L).
```

```
L = [a,b,c,d] ? ;
```

```
no
```

```
| ?- append(X,Y,L).
```

```
X = [],
```

```
Y = L ? ;
```

```
L = [_A|Y],
```

```
X = [_A] ? ;
```

```
L = [_A,_B|Y],
```

```
X = [_A,_B] ?
```

```
yes
```

Reversing a list

```
reverse([], []).
reverse([X|L],R):-reverse(L,K),append(K,[X],R).
| ?- reverse([a,b,c,d],M).
M = [d,c,b,a] ? ;
no
| ?- reverse(M,[a,b,c,d]).
M = [d,c,b,a] ?

rev(L,R):-rev_lin(L,[],R).
rev_lin([],R,R).
rev_lin([X|L],K,R):-rev_lin(L,[X|K],R).
| ?- reverse(X,Y).
X = [], Y = [] ? ;
X = [_A], Y = [_A] ? ;
...
```


Quicksort

```
quicksort([], []).
quicksort([X|L], R):-
    partition(L, Linf, X, Lsup),
    quicksort(Linf, L1),
    quicksort(Lsup, L2),
    append(L1, [X|L2], R).
partition([], [], _, []).
partition([Y|L], [Y|Linf], X, Lsup):-
    Y<X,
    partition(L, Linf, X, Lsup).
partition([Y|L], Linf, X, [Y|Lsup]):-
    Y>X,
    partition(L, Linf, X, Lsup).
```

Parsing

```
sentence(L):-nounphrase(L1), verbphrase(L2), append(L1,L2,L).
```

```
nounphrase(L):-determiner(L1), noun(L2), append(L1,L2,L).
```

```
nounphrase(L):-noun(L).
```

```
verbphrase(L):-verb(L).
```

```
verbphrase(L):-verb(L1), nounphrase(L2), append(L1,L2,L).
```

```
verb([eats]).
```

```
determiner([the]).
```

```
noun([monkey]).
```

```
noun([banana]).
```

Parsing/Synthesis (continued)

```
| ?- sentence([the,monkey,eats]).
```

```
yes
```

```
| ?- sentence([the,eats]).
```

```
no
```

```
| ?- sentence(L).
```

```
L = [the,monkey,eats] ? ;
```

```
L = [the,monkey,eats,the,monkey] ? ;
```

```
L = [the,monkey,eats,the,banana] ? ;
```

```
L = [the,monkey,eats,monkey] ?
```

```
yes
```

Prolog Meta-interpreter

```

solve((A,B)) :- solve(A), solve(B).
solve(A) :- clause(A).
solve(A) :- clause((A:-B)), solve(B).

clause(member(X, [X|_])).
clause((member(X, [_|L]) :- member(X,L))).

| ?- solve(member(X,L)).

L = [X|_A] ? ;
L = [_A,X|_B] ? ;
L = [_A,_B,X|_C] ? ;
L = [_A,_B,_C,X|_D] ?
yes
    
```

Linear Programming

- Variables with a **continuous domain** \mathbb{R} .

$$A.x \leq B \quad \max c.x$$

Satisfiability and optimization has **polynomial complexity** (Simplex algorithm, interior point method).

- Mixed Integer Linear Programming
Variables with either a continuous domain \mathbb{R} or a **discrete domain** \mathbb{Z}

$$x \in \mathbb{Z} \quad A.x \leq B \quad \max c.x$$

NP-hard problem (Branch and bound procedure, Gomory's cuts,...)

CLP(\mathcal{R}) mortgage program

```

int(P,T,I,B,M):- T > 0, T <= 1, B + M = P * (1 + I).
int(P,T,I,B,M):- T > 1, int(P * (1 + I) - M, T - 1, I, B, M).

| ?- int(120000,120,0.01,0,M).
M = 1721.651381 ?
yes
| ?- int(P,120,0.01,0,1721.651381).
P = 120000 ?
yes
| ?- int(P,120,0.01,0,M).
P = 69.700522*M ?
yes
| ?- int(P,120,0.01,B,M).
P = 0.302995*B + 69.700522*M ?
yes
| ?- int(999, 3, Int, 0, 400).
400 = (-400 + (599 + 999*Int) * (1 + Int)) * (1 + Int) ?

```

CLP(\mathcal{R}) heat equation

```

| ?- X=[[0,0,0,0,0,0,0,0,0,0,0,0],
        [100,_,_,_,_,_,_,_,_,100],
        [100,_,_,_,_,_,_,_,_,100],
        [100,_,_,_,_,_,_,_,_,100],
        [100,_,_,_,_,_,_,_,_,100],
        [100,_,_,_,_,_,_,_,_,100],
        [100,_,_,_,_,_,_,_,_,100],
        [100,_,_,_,_,_,_,_,_,100],
        [100,_,_,_,_,_,_,_,_,100],
        [100,_,_,_,_,_,_,_,_,100],
        [100,100,100,100,100,100,100,100,100,100,100]],
    laplace(X).

X=[[0,0,0,0,0,0,0,0,0,0,0,0],
   [100,51.11,32.52,24.56,21.11,20.12,21.11,24.56,32.52,51.11,100],
   [100,71.91,54.41,44.63,39.74,38.26,39.74,44.63,54.41,71.91,100],
   [100,82.12,68.59,59.80,54.97,53.44,54.97,59.80,68.59,82.12,100],
   [100,87.97,78.03,71.00,66.90,65.56,66.90,71.00,78.03,87.97,100],
   [100,91.71,84.58,79.28,76.07,75.00,76.07,79.28,84.58,91.71,100],
   [100,94.30,89.29,85.47,83.10,82.30,83.10,85.47,89.29,94.30,100],
   [100,96.20,92.82,90.20,88.56,88.00,88.56,90.20,92.82,96.20,100],
   [100,97.67,95.59,93.96,92.93,92.58,92.93,93.96,95.59,97.67,100],
   [100,98.89,97.90,97.12,96.63,96.46,96.63,97.12,97.90,98.89,100],
   [100,100,100,100,100,100,100,100,100,100,100]] ?
    
```

CLP(\mathcal{R}) heat equation

```
laplace([H1,H2,H3|T]):-laplace_vec(H1,H2,H3),laplace([H2,H3|T]).
laplace([_,_]).
```

```
laplace_vec([TL,T,TR|T1],[ML,M,MR|T2],[BL,B,BR|T3]):-
    B + T + ML + MR - 4 * M = 0,
    laplace_vec([T,TR|T1],[M,MR|T2],[B,BR|T3]).
```

```
laplace_vec([_,_],[_,_],[_,_]).
```

```
| ?- laplace([[B11, B12, B13, B14],
              [B21, M22, M23, B24],
              [B31, M32, M33, B34],
              [B41, B42, B43, B44]]).
```

B12 = -B21 - 4*B31 + 16*M32 - 8*M33 + B34 - 4*B42 + B43,

B13 = -B24 + B31 - 8*M32 + 16*M33 - 4*B34 + B42 - 4*B43,

M22 = -B31 + 4*M32 - M33 - B42,

M23 = -M32 + 4*M33 - B34 - B43 ?

CLP(\mathcal{FD}) = over Finite Domains

Variables $\{x_1, \dots, x_v\}$

over a **finite domain** $D = \{e_1, \dots, e_d\}$.

Constraints to satisfy:

- unary constraints of domains $x \in \{e_i, e_j, e_k\}$
- binary constraints: $c(x, y)$
 defined intentionally, $x > y + 2$,
 or extensionally, $\{c(a, b), c(d, c), c(a, d)\}$.
- n-ary *global constraints*: $c(x_1, \dots, x_n)$,

CLP(\mathcal{FD}) N-Queens Problem

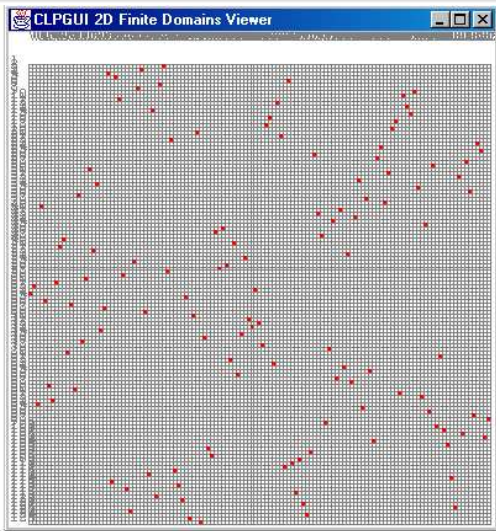
GNU-Prolog program:

```
queens(N,L):-  
    length(L,N),  
    fd_domain(L,1,N),  
    safe(L),  
    fd_labeling(L,first_fail).  
safe([]).  
safe([X|L]):-  
    noattack(L,X,1),  
    safe(L).  
noattack([],_,_).  
noattack([Y|L],X,I):-  
    X#\=Y,  
    X#\=Y+I,  
    X+I#\=Y,  
    I1 is I+1,  
    noattack(L,X,I1).
```

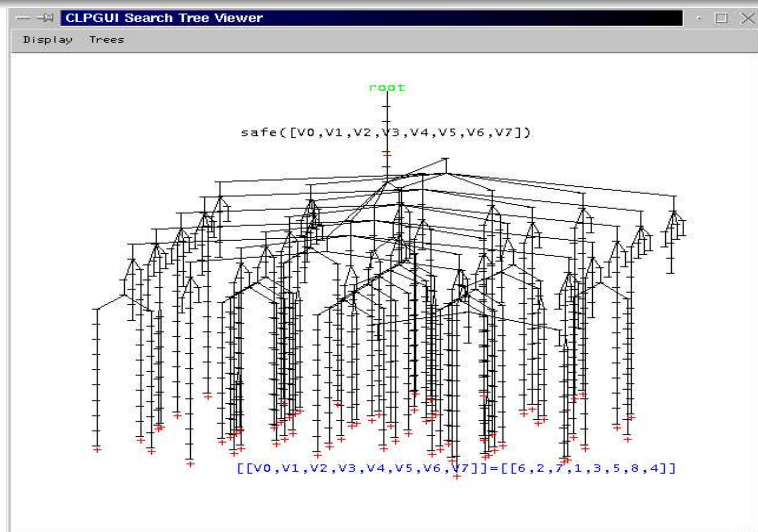
CLP(\mathcal{FD}) N-Queens Problem

GNU-Prolog program:

```
queens(N,L):-
    length(L,N),
    fd_domain(L,1,N),
    safe(L),
    fd_labeling(L,first_fail).
safe([]).
safe([X|L]):-
    noattack(L,X,1),
    safe(L).
noattack([],_,_).
noattack([Y|L],X,I):-
    X#\=Y,
    X#\=Y+I,
    X+I#\=Y,
    I1 is I+1,
    noattack(L,X,I1).
```



Search space of all solutions



CLP(\mathcal{FD}) send+more=money

```

send(L):-sendc(L), labeling(L).
sendc([S,E,N,D,M,O,R,Y]) :-
    domain([S,E,N,D,M,O,R,Y],[0,9]),
    alldifferent([S,E,N,D,M,O,R,Y]), S#/=0, M#/=0,
    eqln(    1000*S+100*E+10*N+D
           + 1000*M+100*O+10*R+E ,
           10000*M+1000*O+100*N+10*E+Y).

| ?- send(L).
L = [9,5,6,7,1,0,8,2] ? ;
no
  
```

CLP(\mathcal{FD}) send+more=money

```
| ?- send([S,E,N,D,M,O,R,Y]).
```

```
D = 1,
```

```
O = 0,
```

```
S = 9,
```

```
domain(E, [4,7]),
```

```
domain(N, [5,8]),
```

```
domain(D, [2,8]),
```

```
domain(R, [2,8]),
```

```
domain(Y, [2,8]),
```

```
Y+90*N#=10*R+D+91*E,
```

```
alldifferent([E,N,D,R,Y]) ?
```

CLP(\mathcal{FD}) scheduling

Simple PERT problem

```
| ?- minimise((B#>=A+5,C#>=B+2,D#>=B+3,E#>=C+5,E#>=D+5), E).
```

Solution de cout 13

```
A = 0, B = 5, D = 8, E = 13, domain(C, [7,8]) ?
```

yes

Disjunctive scheduling is NP-hard

```
| ?- minimise((B#>=A+5,C#>=B+2,D#>=B+3,E#>=C+5,
               E#>=D+5, (C#>=D+5 ; D#>=C+5)), E).
```

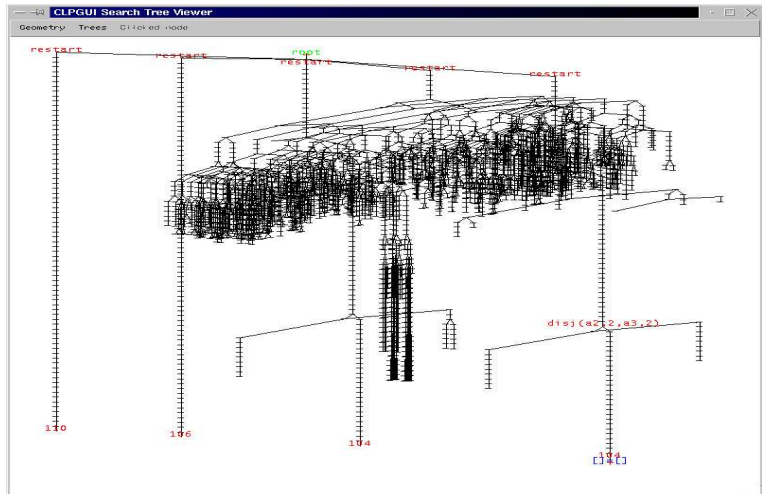
Solution de cout 18

Solution de cout 17

```
A = 0, B = 5, C = 7, D = 12, E = 17 ? ;
```

no

Disjunctive scheduling: bridge problem (4000 nodes)



Reified constraints in CLP($\mathcal{B}, \mathcal{FD}$)

The **reified constraint** $B \Leftrightarrow (X < Y)$
 associates a **boolean** variable B to the satisfaction of the
 constraint $X < Y$

Cardinality constraint $card(N, [C_1, \dots, C_m])$ is true iff there are
 exactly N constraints true in $[C_1, \dots, C_m]$.

```
card(0, []).
card(N, [C|L]) :-
    fd_domain_bool(B),
    B #<=> C,
    N #= B+M,
    card(M, L).
```

Magic Series

Find a sequence of integers (i_0, \dots, i_{n-1}) such that i_j is the number of occurrences of the integer j in the sequence

$$\bigwedge_{j=0}^{n-1} \text{card}(i_j, [i_0 = j, \dots, i_{n-1} = j])$$

Constraint propagation with reified constraints $b_k \Leftrightarrow i_k = j$,
Redundant constraints $n = \sum_{j=0}^{n-1} i_j$,
Enumeration with first fail heuristics,
Less than one second CPU for $n = 50\dots$

Multiple Modeling in CLP(\mathcal{FD})

N-queens with two **concurrent models**: by lines and by columns

```
queens2(N,L) :-  
    list(N, Column), fd_domain(Column,1,N), safe(Column),  
    list(N, Line), fd_domain(Line,1,N), safe(Line),  
    linking(Line,1,Column),  
    append(Line,Column,L), fd_labelingff(L).  
linking([],_,_).  
linking([X|L],I,C):- equivalence(X,I,C,1),  
    I1 is I+1,  
    linking(L,I1,C).  
equivalence(_,_,[],_).  
equivalence(X,I,[Y|L],J):- B#<=>(X#=J), B#<=>(Y#=I),  
    J1 is J+1,  
    equivalence(X,I,L,J1).
```

Programming in CLP($\mathcal{H}, \mathcal{B}, \mathcal{FD}, \mathcal{R}$)

- **Basic constraints** on domains of terms \mathcal{H} , bounded integers \mathcal{FD} , reals \mathcal{R} , booleans \mathcal{B} , ontologies \mathcal{H}_{\leq} , etc.

- **Relations defined extensionally** by constrained facts:

precedence(X, D, Y) :- $X + D < Y$.

disjonctives(X, D, Y, E) :- $X + D < Y$.

disjonctives(X, D, Y, E) :- $Y + E < X$.

and *intentionally* by rules:

labeling($[]$).

labeling($[X|L]$) :- indomain(X), labeling(L).

- Programming of search procedures and heuristics:

And-parallelism (variable choice): “first-fail” heuristics min domain

Or-parallelism (value choice): “best-first” heuristics min value