

Constraint Logic Programming (CLP): a short tutorial

- What is CLP?
 - the use of a rich and powerful language to model optimization problems
 - modelling based on variables, domains and constraints

CLP

- Motivation:

1. to offer a declarative way of modelling constraint satisfaction problems (CSP)
2. to solve 2 limitations in Prolog:
 - Each term in Prolog needs to be explicitly evaluated and is not interpreted (evaluated):
 - $X + 1$ is a term that is not evaluated in Prolog. It is only a syntactic representation.
 - a variable can assume one single value.
 - Uniform computation, but not that powerful: depth-first search, “generate-and-test”.
3. to integrate Artificial Intelligence (AI) and Operations Research (OR)

CLP

- CLP can use Artificial Intelligence (AI) techniques to improve the search: propagation, data-driven computation, “forward checking” and “lookahead”.
- Applications: planning, scheduling, resource allocation, computer graphics, digital circuit design, fault diagnosis etc.
- Clients: Michelin and Dassault, French railway SNCF, Swissair, SAS and Cathay Pacific, HK International terminals, Eriksson, British Telecom etc.

CLP

- CLP joins 2 research areas:
 - Introduction of richer and powerful data structures to Logic Programming (e.g.: replace unification by efficient manipulation of constraints and domains).
 - Consistency techniques:
“generate-and-test” x “constrain-and-generate”

CLP

- Systems:
 - CHIP, Dincbas and Van Hentenryck (ECRC)
 - OPL, Van Hentenryck
 - Xpress
 - CLP(R), Jaffar, Michaylov, Stuckey and Yap (Monash)
 - Prolog III, Colmerauer
 - ECLiPSe, Wallace (IC Parc)
 - Oz, Smolka (DFKI)
 - clp(FD), Diaz and Codognet (INRIA, France)
- The CLP(X) scheme:
 - “constraint solver”: replaces simple unification.
 - 2 popular domains: arithmetic and boolean.

Arithmetic domain: linear constraints

- Prolog can not solve $x-3 = y+5$.
- CLP(R): first language to introduce arithmetic constraints.
- Linear arithmetic expressions composed by: numbers, variables and operators (negation, addition, subtraction, multiplication and division).
- Example: $t1 R t2$, with $R = \{ >, \geq, =, \leq, <, = \}$
- Popular decision procedures:
 - Gauss elimination.
 - Simplex (most popular):
 - average good behavior
 - popular
 - incremental

Arithmetic Domain: Linear Constraints

- Example:
- A meal consists of starter, main course and dessert
- database with various kinds of food and their caloric values
- Problem: produce a menu with light meals (caloric value $< 10\text{Kcal}$)

Arithmetic Domain: Linear Constraints

```
light_meal(A,M,D) :-
    I > 0, J > 0, K > 0,
    I + J + K =< 10,
    starter(A,I),
    main_course(M,J),
    dessert(D,K).

main_course(M,I) :-
    meat(M,I).
main_course(M,I) :-
    fish(M,I).
starter(salad,1).
starter(soup,6).

meat(steak,5).
meat(pork,7).
fish(sole,2).
fish(tuna,4).
dessert(fruit,2).
dessert(icecream,6).
```

Arithmetic Domain: Linear Constraints

- Intermediate results = compute states.
- 2 components: constraint store and continuation of objectives.
- Query: \diamond `light_meal(A,M,D)`.
- $I + J + K \leq 10, I > 0, J > 0, K > 0 \diamond$ `starter(A,I), main_course(M,J), dessert(D,K)`.
- $A = \text{salad}, I = 1, 1 + J + K \leq 10, 1 > 0, J > 0, K > 0 \diamond$ `main_course(M,J), dessert(D,K)`.
- $A = \text{salad}, I = 1, M=M1, J=I1, 1 + J + K \leq 10, 1 > 0, J > 0, K > 0 \diamond$ `meat(M1,I1), dessert(D,K)`.
- $A = \text{salad}, I = 1, M=\text{steak}, J=5, M1=\text{steak}, I1 = 5, 1 + 5 + K \leq 10, 1 > 0, 5 > 0, K > 0 \diamond$ `dessert(D,K)`.
- $A = \text{salad}, I = 1, M=\text{steak}, J=5, M1=\text{steak}, I1 = 5, D=\text{fruit}, K = 2, 1 + 5 + 2 \leq 10, 1 > 0, 5 > 0, 2 > 0 \diamond$.

Arithmetic Domain: Linear Constraints

Inconsistent derivation :

- $A = \text{pasta}, I = 6, M = \text{steak}, J = 5, M1 = \text{steak}, I1 = 5, 6 + 5 + K$
 $= \langle 10, 5 \rangle, 6 > 0, K > 0 \diamond \text{dessert}(D, K).$

Arithmetic Domain: non-linear Constraints

- Example: multiply 2 complex numbers: $(R1 + I*I1) * (R2 + I*I2)$.

```
zmul(R1,I1,R2,I2,R3,I3) :-  
    R3 = R1 * R2 + I1 * I2,  
    I3 = R1 * I2 + R2 * I1.
```

- Query: \diamond `zmul(1,2,3,4,R3,I3)`

- Equations become linear.

- Solution: $R3 = -5, I3 = 10$ (definite solution)

- Query: \diamond `zmul(1,2,R2,I2,R3,I3)`

- Solution:

```
I2 = 0.2*I3 - 0.4*R3  
R2 = 0.4*I3 + 0.2*R3  
yes (undefined solution)
```

Arithmetic Domain: non-linear Constraints

- Same example: multiply 2 complex numbers:

$$(R1 + I*I1) * (R2 + I*I2)$$

- Query: $\diamond \text{zmul}(R1, 2, R2, 4, -5, 10), R2 < 3$.

- CLP(R): (do not solve non-linear equations)

$$R1 = -0.5*R2 + 2.5$$

$$3 = R1*R2$$

$$R2 < 3$$

Maybe

- applications of non-linear equations: computational geometry and financial applications (various algorithms used).

Boolean domain

- Main Application: digital circuit design (hardware verification) and theorem proof.
- Boolean terms: truth values (F or T), variables, logical operators, one single constraint: equality.
- various unification algorithms for boolean constraints.
- Solution: provides a decision procedure for propositional calculus (NP-complete).

Boolean domain

- Example: full adder (operators # (xor), * (and), + (or))

add(I1,I2,I3,01,02) :-

$$X1 = I1 \# I2,$$

$$A1 = I1 * I2,$$

$$O1 = X1 \# I3,$$

$$A2 = I3 * X1,$$

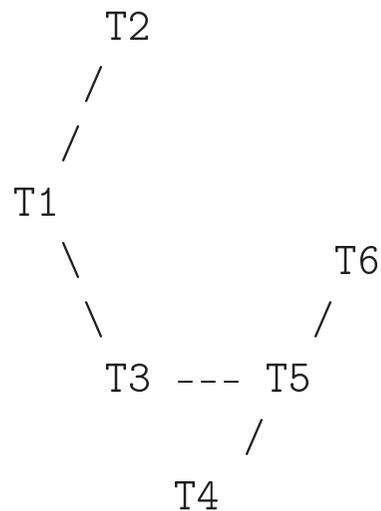
$$O2 = A1 + A2.$$

- Query: \diamond add(a,b,c,01,02)

- Solution: $O1 = a+b+c$, $O2 = (a * b) + (a * c) \# (b * c)$

Consistency techniques

- Eliminate inconsistent 'labellings' by constraint propagation information about the values of variables.
- examples: arc-consistency, forward checking, generalized propagation.
- Example: task scheduling.



```
before(T1,T2) .  
before(T1,T3) .  
before(T2,T6) .  
before(T3,T5) .  
before(T4,T5) .  
before(T5,T6) .  
notequal(T2,T3) .
```

Consistency Techniques

- Example: $T1 \in \{1, 2, 3, 4, 5\}$, $T2 \in \{1, 2, 3, 4, 5\}$.
- $\text{before}(T1, T2) \rightarrow$ apply consistency:
 - $T1 \in \{1, 2, 3, 4\}$
 - $T2 \in \{2, 3, 4, 5\}$
- Value 5 removed from T1, because there is no other value in T2 that can satisfy $T1 < T2$.
- Value 1 removed from T2, for the same reason.

Arc Consistency

- $T1 \in \{1, 2\}$, $T2 \in \{2, 3, 4\}$, $T3 \in \{2, 3\}$, $T4 \in \{1, 2, 3\}$, $T5 \in \{3, 4\}$, $T6 \in \{4, 5\}$.
- Value 2 from T1 is chosen (T1 is “labelled” with value 2).
- Using propagation: $T2 \in \{3, 4\}$ and $T3 \in \{3\}$
- $T2 \in \{4\}$ from $\text{notequal}(T2, T3)$.
- Finally: $T1 \in \{2\}$, $T2 \in \{4\}$, $T3 \in \{3\}$, $T4 \in \{1, 2, 3\}$, $T5 \in \{4\}$, $T6 \in \{5\}$

Infinite Domains

- Utilization of maximum and minimum values to solve constraints in the linear domain.
- E.g.: x, y, z with domains $[1..10]$ with constraint $2x + 3y + 2 < z$
- Removing inconsistent values:
 - 10 is the largest value for z , then: $2x + 3y < 8$
 - 1 is the smallest possible value for y , therefore: $2x < 5$
 - x can only assume values $\{1, 2\}$
 - $3y < 6, y < 2, y \in \{1\}$
 - $z > 7, z \in \{8, 9, 10\}$

Basic programming techniques

- Define problem variables and their domains.
- Establish the constraints between variables.
- Search for solution.

```
?- [X,Y,Z]::1..10,  
    2 * X + 3 * Y + 2 #< Z,  
    indomain(X), indomain(Y), indomain(Z).
```

Algorithms and other examples

Forward Checking:

$n = 8$

(1) $V_1 = 1 \implies$

- $V_2 = \{3, 4, 5, 6, 7, 8\}$
- $V_3 = \{2, 4, 5, 6, 7, 8\}$
- $V_4 = \{2, 3, 5, 6, 7, 8\}$
- $V_5 = \{2, 3, 4, 6, 7, 8\}$
- $V_6 = \{2, 3, 4, 5, 7, 8\}$
- $V_7 = \{2, 3, 4, 5, 6, 8\}$
- $V_8 = \{2, 3, 4, 5, 6, 7\}$

(2) $V_2 = 3 \implies$

- $V_3 = \{5, 6, 7, 8\}$
- $V_4 = \{2, 6, 7, 8\}$
- $V_5 = \{2, 4, 7, 8\}$
- $V_6 = \{2, 4, 5, 8\}$
- $V_7 = \{2, 4, 5, 6\}$
- $V_8 = \{2, 4, 5, 6, 7\}$

Forward Checking: Example

```
n-queens
n=8
after V1=1
    V2=3
+---+---+---+---+---+---+---+---+
| X |   |   |   |   |   |   |   |
+---+---+---+---+---+---+---+---+
|   |   |   |   |   |   |   |   |
+---+---+---+---+---+---+---+---+
|   | X |   |   |   |   |   |   |
+---+---+---+---+---+---+---+---+
|   |   |   |   |   |   |   |   |
+---+---+---+---+---+---+---+---+
|   |   |   |   |   |   |   |   |
+---+---+---+---+---+---+---+---+
|   |   |   |   |   |   |   |   |
+---+---+---+---+---+---+---+---+
|   |   |   |   |   |   |   |   |
+---+---+---+---+---+---+---+---+
|   |   |   |   |   |   |   |   |
+---+---+---+---+---+---+---+---+
```

Algorithms

- infinite domains: linear programming, simplex, revised simplex, convex hull, Gauss elimination.
- finite domains: forward checking, lookahead, arc-consistency.
- For finite domains, 2 problems:
 - choice of variable:
 - *most-constrained*: smallest domain
 - *most constraining*: mostly constrains domains of other variables
 - choice of a value for a variable:
 - *first-fail* principle.
 - *least constraining*: value that constrains less sets of values of other variables

Heuristics

- most-constrained variable: allows to solve the n-queens problem with n equals 100
- pure forward checking: only solves for n = 30
- least-constraining: solves for n = 1000

References

- A comparative study of eight constraint programming languages, by Antonio Fernandez and Pat Hill, 2000
- Constraint Logic Programming - An Informal Introduction, 1993, by Thom Frühwirth, A. Herold, T. Le Provost, P. Lim, E. Monfroy, M. Wallace, V. Küchenhoff, Thierry Le Provost Pierre Lim
- Constraint logic programming: A survey J. Jaffar, M. J. Maher, 1994