

# Constraint Logic Programming

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# Full abstraction

Let  $F_1(P) = \text{lfp}(T_P^{\mathcal{X}}) = T_P^{\mathcal{X}} \uparrow \omega = \dots T_P^{\mathcal{X}}(T_P^{\mathcal{X}}(\emptyset)) \dots$

Theorem ([JL87])

$$F_1(P) = O_{gs}(P).$$

$F_1(P) \subseteq O_{gs}(P)$  is proved by induction on the powers  $n$  of  $T_P^{\mathcal{X}}$ .  $n = 0$  is trivial. Let  $A\rho \in T_P^{\mathcal{X}} \uparrow n$ , there exists a rule  $(A \leftarrow c | A_1, \dots, A_n) \in P$ , s.t.  $\{A_1\rho, \dots, A_n\rho\} \subseteq T_P^{\mathcal{X}} \uparrow n - 1$  and  $\mathcal{X} \models c\rho$ . By induction  $\{A_1\rho, \dots, A_n\rho\} \subseteq O_{gs}(P)$ . By definition of  $O_{gs}$  we get  $A\rho \in O_{gs}(P)$ .

$O_{gs}(P) \subseteq F_1(P)$  is proved by induction on the length of derivations.

Successes with derivation of length 0 are ground facts in  $T_P^{\mathcal{X}} \uparrow 1$ . Let  $A\rho \in O_{gs}(P)$  with a derivation of length  $n$ . By definition of  $O_{gs}$  there exists  $(A \leftarrow c | A_1, \dots, A_n) \in P$  s.t.  $\{A_1\rho, \dots, A_n\rho\} \subseteq O_{gs}(P)$  and  $\mathcal{X} \models c\rho$ . By induction  $\{A_1\rho, \dots, A_n\rho\} \subseteq F_1(P)$ . Hence by definition of  $T_P^{\mathcal{X}}$  we get  $A\rho \in F_1(P)$ .

## $T_P^{\mathcal{X}}$ and $\mathcal{X}$ models

### Proposition

*$I$  is a  $\mathcal{X}$ -model of  $P$  iff  $I$  is a post-fixed point of  $T_P^{\mathcal{X}}$ ,  $T_P^{\mathcal{X}}(I) \subseteq I$ .*

### Proof.

$I$  is a  $\mathcal{X}$ -model of  $P$ ,

iff for each clause  $A \leftarrow c \mid A_1, \dots, A_n \in P$  and for each  $\mathcal{X}$ -valuation  $\rho$ , if  $\mathcal{X} \models c\rho$  and  $\{A_1\rho, \dots, A_n\rho\} \subseteq I$  then  $A\rho \in I$ ,

iff  $T_P^{\mathcal{X}}(I) \subseteq I$ . □

Relating  $S_P^X$  and  $T_P^X$  operators

## Theorem ([JL87])

For every ordinal  $\alpha$ ,  $T_P^X \uparrow \alpha = [S_P^X \uparrow \alpha]_X$ .

## Proof.

The base case  $\alpha = 0$  is trivial. For a successor ordinal, we have

$$\begin{aligned} [S_P^X \uparrow \alpha]_X &= [S_P^X (S_P^X \uparrow \alpha - 1)]_X \\ &= T_P^X ([S_P^X \uparrow \alpha - 1]_X) \\ &= T_P^X (T_P^X \uparrow \alpha - 1) \text{ by induction} \\ &= T_P^X \uparrow \alpha. \end{aligned}$$

For a limit ordinal, we have

$$\begin{aligned} [S_P^X \uparrow \alpha]_X &= [\bigcup_{\beta < \alpha} S_P^X \uparrow \beta]_X \\ &= \bigcup_{\beta < \alpha} [S_P^X \uparrow \beta]_X \\ &= \bigcup_{\beta < \alpha} T_P^X \uparrow \beta \text{ by induction} \\ &= T_P^X \uparrow \alpha \end{aligned}$$



## Full abstraction w.r.t. computed constraints

Theorem (Theorem of full abstraction [GL91])

$$O_{ca}(P) = F_2(P).$$

$F_2(P) \subseteq O_{ca}(P)$  is proved by induction on the powers  $n$  of  $S_P^{\mathcal{X}}$ .  $n = 0$  is trivial. Let  $c|A \in S_P^{\mathcal{X}} \uparrow n$ , there exists a rule  $(A \leftarrow d|A_1, \dots, A_n) \in P$ , s.t.  $\{c_1|A_1, \dots, c_n|A_n\} \subseteq S_P^{\mathcal{X}} \uparrow n - 1$ ,  $c = d \wedge \bigwedge_{i=1}^n c_i$  and  $\mathcal{X} \models \exists c$ . By induction  $\{c_1|A_1, \dots, c_n|A_n\} \subseteq O_{ca}(P)$ . By definition of  $O_{ca}$  we get  $c|A \in O_{ca}(P)$ .

$O_{ca}(P) \subseteq F_2(P)$  is proved by induction on the length of derivations. Successes with derivation of length 0 are facts in  $S_P^{\mathcal{X}} \uparrow 1$ . Let  $c|A \in O_{ca}(P)$  with a derivation of length  $n$ . By definition of  $O_{ca}$  there exists  $(A \leftarrow d|A_1, \dots, A_n) \in P$  s.t.  $\{c_1|A_1, \dots, c_n|A_n\} \subseteq O_{ca}(P)$ ,  $c = d \wedge \bigwedge_{i=1}^n c_i$  and  $\mathcal{X} \models \exists c$ . By induction  $\{c_1|A_1, \dots, c_n|A_n\} \subseteq F_2(P)$ . Hence by definition of  $S_P^{\mathcal{X}}$  we get  $c|A \in F_2(P)$ .



## Part IV

# Logical Semantics

## Part IV: Logical Semantics

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## Logical Semantics of CLP( $\mathcal{X}$ ) Programs

- Proper logical semantics

$$(1) \quad P, \mathcal{T} \models \exists(G) \qquad (4) \quad P, \mathcal{T} \models c \supset G,$$

- Logical semantics in a fixed pre-interpretation

$$(2) \quad P \models_{\mathcal{X}} \exists(G) \qquad (5) \quad P \models_{\mathcal{X}} c \supset G,$$

- Algebraic semantics

$$(3) \quad M_P^{\mathcal{X}} \models \exists(G) \qquad (6) \quad M_P^{\mathcal{X}} \models c \supset G.$$

We show  $(1) \Leftrightarrow (2) \Leftrightarrow (3)$  and  $(4) \Rightarrow (5) \Leftrightarrow (6)$ .

## Soundness of CSLD Resolution

### Theorem ([JL87])

*If  $c$  is a computed answer for the goal  $G$  then  $M_P^{\mathcal{X}} \models c \supset G$ ,  $P \models_{\mathcal{X}} c \supset G$  and  $P, \mathcal{T} \models c \supset G$ .*

If  $G = (d | A_1, \dots, A_n)$ , we deduce from the  $\wedge$ -compositionality lemma, that there exist computed answers  $c_1, \dots, c_n$  for the goals  $A_1, \dots, A_n$  such that  $c = d \wedge \bigwedge_{i=1}^n c_i$  is satisfiable. For every  $1 \leq i \leq n$   $c_i | A_i \in S_P^{\mathcal{X}} \uparrow \omega$ , by the full abstraction Thm, 4,  $[c_i | A_i]_{\mathcal{X}} \subseteq M_P^{\mathcal{X}}$ , by Thm. 3, and Prop. 2, hence  $M_P^{\mathcal{X}} \models \forall (c_i \supset A_i)$ ,  $P \models_{\mathcal{X}} \forall (c_i \supset A_i)$  as  $M_P^{\mathcal{X}}$  is the least  $\mathcal{X}$ -model of  $P$ ,  $P \models_{\mathcal{X}} \forall (c \supset A_i)$  as  $\mathcal{X} \models \forall (c \supset c_i)$  for all  $i$ ,  $1 \leq i \leq n$ . Therefore we have  $P \models_{\mathcal{X}} \forall (c \supset (d \wedge A_1 \wedge \dots \wedge A_n))$ , and as the same reasoning applies to any model  $\mathcal{X}$  of  $\mathcal{T}$ ,  $P, \mathcal{T} \models \forall (c \supset (d \wedge A_1 \wedge \dots \wedge A_n))$

# Completeness of CSLD resolution

## Theorem ([Mah87])

*If  $M_P^{\mathcal{X}} \models_{\mathcal{X}} c \supset G$  then there exists a set  $\{c_i\}_{i \geq 0}$  of computed answers for  $G$ , such that:  $\mathcal{X} \models \forall (c \supset \bigvee_{i \geq 0} \exists Y_i c_i)$ .*

## Proof.

For every solution  $\rho$  of  $c$ , for every atom  $A_j$  in  $G$ ,

$M_P^{\mathcal{X}} \models A_j \rho$  iff  $A_j \rho \in T_P^{\mathcal{X}} \uparrow \omega$ , by Thm. 1, iff  $A_j \rho \in [S_P^{\mathcal{X}} \uparrow \omega]_{\mathcal{X}}$ , by Thm. 3,

iff  $c_{j,\rho} | A_j \in S_P^{\mathcal{X}} \uparrow \omega$ , for some constraint  $c_{j,\rho}$  s.t.  $\rho$  is solution of  $\exists Y_{j,\rho} c_{j,\rho}$ , where  $Y_{j,\rho} = V(c_{j,\rho}) \setminus V(A_j)$ ,

iff  $c_{j,\rho}$  is a computed answer for  $A_j$  (by 4) and  $\mathcal{X} \models \exists Y_{j,\rho} c_{j,\rho}$ .

Let  $c_{\rho}$  be the conjunction of  $c_{j,\rho}$  for all  $j$ .  $c_{\rho}$  is a computed answer for  $G$ .

By taking the collection of  $c_{\rho}$  for all  $\rho$  we get  $\mathcal{X} \models \forall (c \supset \bigvee_{c_{\rho}} \exists Y_{\rho} c_{\rho})$

□

## Completeness w.r.t. the theory of the structure

### Theorem ([Mah87])

If  $P, \mathcal{T} \models c \supset G$  then there exists a finite set  $\{c_1, \dots, c_n\}$  of computed answers to  $G$ , such that:  
 $\mathcal{T} \models \forall (c \supset \exists Y_1 c_1 \vee \dots \vee \exists Y_n c_n)$ .

### Proof.

If  $P, \mathcal{T} \models c \supset G$  then for every model  $\mathcal{X}$  of  $\mathcal{T}$ , for every  $\mathcal{X}$ -solution  $\rho$  of  $c$ , there exists a computed constraint  $c_{\mathcal{X}, \rho}$  for  $G$  s.t.  $\mathcal{X} \models c_{\mathcal{X}, \rho} \rho$ . Let  $\{c_i\}_{i \geq 0}$  be the set of these computed answers. Then for every model  $\mathcal{X}$  and for every  $\mathcal{X}$ -valuation  $\rho$ ,  $\mathcal{X} \models c \supset \bigvee_{i \geq 1} \exists Y_i c_i$ , therefore

$$\mathcal{T} \models c \supset \bigvee_{i \geq 1} \exists Y_i c_i,$$

As  $\mathcal{T} \cup \{\exists (c \wedge \neg \exists Y_i c_i)\}_i$  is unsatisfiable, by applying the compactness theorem of first-order logic there exists a finite part  $\{c_i\}_{1 \leq i \leq n}$ ,

$$\text{s.t. } \mathcal{T} \models c \supset \bigvee_{i=1}^n \exists Y_i c_i.$$



## First-order theorem proving in CLP( $\mathcal{H}$ )

Prolog can be used to find proofs by refutation of Horn clauses (with a complete search meta-interpreter).

$P, \forall(\neg A)$  is unsatisfiable iff  $P \models \exists(A)$  iff  $A \longrightarrow^* \square$ .

Groups can be axiomatized with Horn clauses with a ternary predicate  $p(x, y, z)$  meaning  $x * y = z$ .

```

clause(p(e,X,X)).
clause(p(i(X),X,e)).
clause((p(U,Z,W) :- p(X,Y,U), p(Y,Z,V), p(X,V,W))).
clause((p(X,V,W) :- p(X,Y,U), p(Y,Z,V), p(U,Z,W))).
    
```

## Theorem proving in groups

To show  $i(i(x)) = x$  by refutation,  
 we show that the formula  $\neg \forall x p(i(i(X)), e, X)$  is unsatisfiable  
 By Skolemization we get the goal clause  $\neg p(i(i(a)), e, a)$

```
| ?- solve(p(i(i(a)),e,a)).
```

```
depth 2
```

```
yes
```

```
| ?- solve(p(a,e,a)).
```

```
depth 4
```

```
yes
```

```
| ?- solve(p(a,i(a),e)).
```

```
depth 3
```

```
yes
```



## Theorem proving in groups (cont.)

To show that any non empty subset of a group, stable by division, is a subgroup we add two clauses

```
clause(s(a)).
```

```
clause((s(Z) :- s(X), s(Y), p(X,i(Y),Z))).
```

and prove that  $s$  contains  $e$  and  $i(a)$ .

```
| ?- solve(s(e)).
```

```
depth 4
```

```
yes
```

```
| ?- solve(s(i(a))).
```

```
depth 5
```

```
yes
```

## Higher-order theorem proving in CLP( $\lambda$ )

Church's simply typed  $\lambda$ -calculus

$$t ::= v \mid t_1 \rightarrow t_2$$

$$e : t ::= x : t \mid (\lambda x : t_1. e : t_2) : t_1 \rightarrow t_2 \mid (e_1 : t_1 \rightarrow t_2(e_2 : t_1)) : t_2$$

Theory of functionality

$$\lambda x. e_1 =_{\alpha} \lambda y. e_1[y/x] \text{ if } y \notin V(e_1),$$

$$(\lambda x. e_1)e_2 \rightarrow_{\beta} e_1[e_2/x]$$

$=_{\alpha} . \rightarrow_{\beta}$  is terminating and confluent

$$e_1 =_{\alpha, \beta} e_2 \text{ iff } \downarrow_{\beta} e_1 =_{\alpha} \downarrow_{\beta} e_2.$$

Equality is decidable, but not unification...

# Theorem proving in CLP( $\lambda$ )

## Theorem (Cantor's Theorem)

$\mathbb{N}^{\mathbb{N}}$  is not countable.

### Proof.

By two steps of CSLD resolution!

Let us suppose  $\exists h : \mathbb{N} \rightarrow (\mathbb{N} \rightarrow \mathbb{N}) \forall f : \mathbb{N} \rightarrow \mathbb{N} \exists n : \mathbb{N} h(n) = f$

After Skolemisation we get  $\forall F h(n(F)) = F$ , i.e.  $\forall F \neg h(n(F)) \neq F$ .

Let us consider the following program  $G \neq H \leftarrow G(N) \neq H(N).$   
 $N \neq s(N).$

We have  $h(n F) \neq F \xrightarrow{\sigma_1} (h(n F))(I) \neq F(I) \xrightarrow{\sigma_2} \square$

where the unifier  $\sigma_2 = \{G = h \mid I, I = n(F), F = \lambda i.s(h \ i \ i), H = F\}$

is Cantor's diagonal argument! □

## Negation as Failure

A **derivation CSLD is fair** if every atom which appears in a goal of the derivation is selected after a finite number of resolution steps.

A **fair CSLD tree** for a goal  $G$  is a CSLD derivation tree for  $G$  in which all derivations are fair.

A goal  $G$  is **finitely failed** if  $G$  has a fair CSLD derivation tree to  $G$ , which is finite and which contains no success.

$p \text{ :- } p.$

$| \text{ ?- member}(a, [b, c, d]).$

no

$| \text{ ?- } p, \text{ member}(a, [b, c, d]).$

...

## Logical semantics of finite failure?

Horn clauses entail no negative information: the Herbrand's base  $\mathcal{B}_{\mathcal{X}}$  is a model.

On the other hand, **the complement of the least  $\mathcal{X}$ -model  $M_P^{\mathcal{X}}$  is not recursively enumerable.**

Indeed let us suppose the opposite. We could define in Prolog the predicates:

- `success(P,B)` which succeeds iff  $M_P \models \exists B$ , i.e. if the goal  $B$  has a successful SLD derivation with the program  $P$
- `fail(P,B)` which succeeds iff  $M_P \models \neg \exists B$

## Undecidability of $M_P^{\mathcal{X}}$

```
loop:- loop.  
contr(P):- success(P,P), loop.  
contr(P):- fail(P,P).
```

If `contr(contr)` has a success,  
then `success(contr,contr)` succeeds,  
and `fail(contr,contr)` doesn't succeed,  
hence `contr(contr)` doesn't succeed: contradiction.

If `contr(contr)` doesn't succeed,  
then `fail(contr,contr)` succeeds,  
hence `contr(contr)` succeeds: contradiction.

Therefore **programs success and fail cannot exist.**

## Clark's completion

The **Clark's completion** of  $P$  is the set  $P^*$  of formulas of the form  $\forall X p(X) \leftrightarrow (\exists Y_1 c_1 \wedge A_1^1 \wedge \dots \wedge A_{n_1}^1) \vee \dots \vee (\exists Y_k c_k \wedge A_1^k \wedge \dots \wedge A_{n_k}^k)$  where the  $p(X) \leftarrow c_i | A_1^i, \dots, A_{n_i}^i$  are the rules in  $P$  and  $Y_i$ 's the local variables,  
 $\forall X \neg p(X)$  if  $p$  is not defined in  $P$ .

### Example

CLP( $\mathcal{H}$ ) program  $p(s(X)) :- p(X)$ .

Clark's completion  $P^* = \{\forall x p(x) \leftrightarrow \exists y x = s(y) \wedge p(y)\}$ .

The goal  $p(0)$  finitely fails, we have  $P^*, CET \models \neg p(0)$ .

The goal  $p(X)$  doesn't finitely fail,

we have  $P^*, CET \not\models \neg \exists X p(X)$  although  $P^* \models_{\mathcal{H}} \neg \exists X p(X)$

## Supported $\mathcal{X}$ -models

### Proposition

*i)  $I$  is a supported  $\mathcal{X}$ -model of  $P$  iff ii)  $I$  is a  $\mathcal{X}$ -model of  $P^*$  iff iii)  $I$  is a fixed point of  $T_P^{\mathcal{X}}$ .*

### Proof.

$I$  is a  $\mathcal{X}$ -model of  $P^*$

iff  $I$  is a  $\mathcal{X}$ -model of  $\forall X p(X) \leftarrow \phi_1 \vee \dots \vee \phi_k$  for every formula  
 $\forall X p(X) \leftrightarrow \phi_1 \vee \dots \vee \phi_k$  in  $P^*$ ,

iff  $I$  is a post-fixed point of  $T_P^{\mathcal{X}}$ , i.e.  $T_P^{\mathcal{X}}(I) \subseteq I$ .

$I$  is a **supported**  $\mathcal{X}$ -interpretation of  $P$ ,

iff  $I$  is a  $\mathcal{X}$ -model of  $\forall X p(X) \rightarrow \phi_1 \vee \dots \vee \phi_k$  for every formula  
 $\forall X p(X) \leftrightarrow \phi_1 \vee \dots \vee \phi_k$  in  $P^*$ ,

iff  $I$  is a pre-fixed point of  $T_P^{\mathcal{X}}$ , i.e.  $I \subseteq T_P^{\mathcal{X}}(I)$ .

Thus *i)  $I$  is a supported  $\mathcal{X}$ -model of  $P$  iff ii)  $I$  is a  $\mathcal{X}$ -model of  $P^*$  iff iii)  $I$  is a fixed point of  $T_P^{\mathcal{X}}$ .* □



## Models of the Clark's completion

### Theorem

- i)  $P^*$  has the same least  $\mathcal{X}$ -model than  $P$ ,  $M_P^{\mathcal{X}} = M_{P^*}^{\mathcal{X}}$
- ii)  $P \models_{\mathcal{X}} c \supset A$  iff  $P^* \models_{\mathcal{X}} c \supset A$ , for all  $c$  and  $A$ ,
- iii)  $P, \mathcal{T} \models c \supset A$  iff  $P^*, \mathcal{T} \models c \supset A$ .

### Proof.

i) is an immediate corollary of full abstraction and least  $\mathcal{X}$ -model theorems.

For iii) we clearly have  $(P, \mathcal{T} \models c \supset A) \Rightarrow (P^*, \mathcal{T} \models c \supset A)$ . We show the contrapositive of the opposite,  $(P, \mathcal{T} \not\models c \supset A) \Rightarrow (P^*, \mathcal{T} \not\models c \supset A)$ .

Let  $I$  be a model of  $P$  and  $\mathcal{T}$ , based on a structure  $\mathcal{X}$ , let  $\rho$  be a valuation such that  $I \models \neg A\rho$  and  $\mathcal{X} \models c\rho$ .

We have  $M_P^{\mathcal{X}} \models \neg A\rho$ , thus  $M_{P^*}^{\mathcal{X}} \models \neg A\rho$ , and as  $\mathcal{T} \models c\rho$ , we conclude that  $P^*, \mathcal{T} \not\models c \supset A$ .

The proof of ii) is identical, the structure  $\mathcal{X}$  being fixed. □

## Soundness of Negation as Finite Failure

### Theorem

*If  $G$  is finitely failed then  $P^*, \mathcal{T} \models \neg G$ .*

### Proof.

By induction on the height  $h$  of the tree in finite failure for  $G = c|A, \alpha$  where  $A$  is the selected atom at the root of the tree.

In the base case  $h = 1$ , the constrained atom  $c|A$  has no CSLD transition, we can deduce that  $P^*, \mathcal{T} \models \neg(c \wedge A)$  hence that  $P^*, \mathcal{T} \models \neg G$ .

For the induction step, let us suppose  $h > 1$ . Let  $G_1, \dots, G_n$  be the sons of the root and  $Y_1, \dots, Y_n$  be the respective sets of introduced variables. We have  $P^*, \mathcal{T} \models G \leftrightarrow \exists Y_1 G_1 \vee \dots \vee \exists Y_n G_n$ . By induction hypothesis,  $P^*, \mathcal{T} \models \neg G_i$  for every  $1 \leq i \leq n$ , therefore  $P^*, \mathcal{T} \models \neg G$ .  $\square$

## Completeness of Negation as Failure

### Theorem ([JL87])

*If  $P^*, \mathcal{T} \models \neg G$  then  $G$  is finitely failed.*

We show that if  $G$  is not finitely failed then  $P^*, \mathcal{T}, \exists(G)$  is satisfiable. If  $G$  has a success then by the soundness of CSLD resolution,  $P^*, \mathcal{T} \models \exists G$ . Else  $G$  has a fair infinite derivation  $G = c_0 | G_0 \longrightarrow c_1 | G_1 \longrightarrow \dots$

For every  $i \geq 0$ ,  $c_i$  is  $\mathcal{T}$ -satisfiable, thus by the **compactness theorem**,  $c_\omega = \bigcup_{i \geq 0} c_i$  is  $\mathcal{T}$ -satisfiable. Let  $\mathcal{X}$  be a model of  $\mathcal{T}$  s.t.  $\mathcal{X} \models \exists(c_\omega)$ .

Let  $l_0 = \{A\rho \mid A \in G_i \text{ for some } i \geq 0 \text{ and } \mathcal{X} \models c_\omega\rho\}$ . As the derivation is fair, every atom  $A$  in  $l_0$  is selected, thus  $c_\omega | A \longrightarrow c_\omega | A_1, \dots, A_n$  with  $[c_\omega | A] \cup \dots \cup [c_\omega | A_n] \subseteq l_0$ . We deduce that  $l_0 \subseteq T_P^{\mathcal{X}}(l_0)$ . By

Knaster-Tarski's theorem, the iterated application up to ordinal  $\omega$  of the operator  $T_P^{\mathcal{X}}$  from  $l_0$  leads to a fixed point  $l$  s.t.  $l_0 \subseteq l$ , thus  $[c_\omega | G_0] \in l$ . Hence  $P^*, \exists(G)$  is  $\mathcal{X}$ -satisfiable, and  $P^*, \mathcal{T}, \exists(G)$  is satisfiable.

# Interlude

## Part V

# Concurrent Constraint Programming

## Part V: Concurrent Constraint Programming

### 17 Introduction

- Syntax
- CC vs. CLP

### 18 Operational Semantics

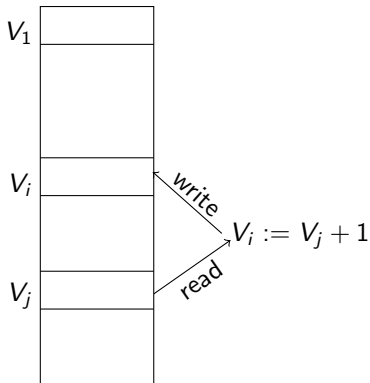
- Transitions
- Properties
- Observables

### 19 Examples

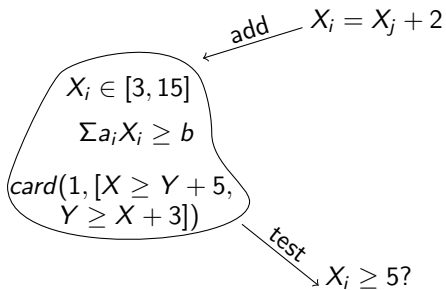
- append
- merge
- $CC(\mathcal{FD})$

# The Paradigm of Constraint Programming

memory of values  
programming variables



memory of constraints  
mathematical variables



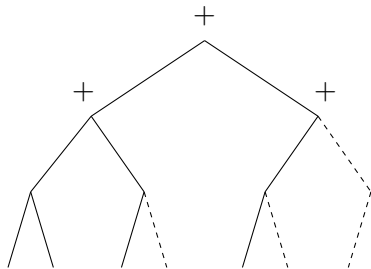
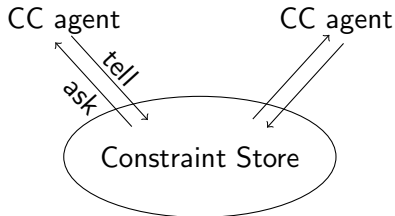
# Concurrent Constraint Programs

Class of programming languages  $CC(\mathcal{X})$  introduced by Saraswat [Sar93] as a merge of Constraint and Concurrent Logic Programming.

Processes  $P ::= D.A$

Declarations  $\mathcal{D} ::= p(\vec{x}) = A, \mathcal{D} \mid \epsilon$

Agents  $A ::= tell(c) \mid \forall \vec{x}(c \rightarrow A) \mid A \parallel A \mid A + A \mid \exists xA \mid p(\vec{x})$





Translating  $\text{CLP}(\mathcal{X})$  into  $\text{CC}(\mathcal{X})$  Declarations

CLP( $\mathcal{X}$ ) program:

$$\begin{aligned} A &\leftarrow c \mid B, C \\ A &\leftarrow d \mid D, E \\ B &\leftarrow e \end{aligned}$$

equivalent  $\text{CC}(\mathcal{X})$  declaration:

$$\begin{aligned} A &= \text{tell}(c) \parallel B \parallel C + \text{tell}(d) \parallel D \parallel E \\ B &= \text{tell}(e) \end{aligned}$$

This is just a **process calculus** syntax for CLP programs. . .

Translating  $CC(\mathcal{X})$  without ask into  $CLP(\mathcal{X})$ 

$(CC \text{ agent})^\dagger = CLP \text{ goal}$

$$(tell(c))^\dagger = c$$

$$(A \parallel B)^\dagger = A^\dagger, B^\dagger$$

$$(A + B)^\dagger = p(\vec{x}) \text{ where } \vec{x} = fv(A) \cup fv(B) \text{ and}$$

$$p(\vec{x}) \leftarrow A^\dagger$$

$$p(\vec{x}) \leftarrow B^\dagger$$

$$(\exists x A)^\dagger = q(\vec{y}) \text{ where } \vec{y} = fv(A) \setminus \{x\} \text{ and}$$

$$q(\vec{y}) \leftarrow A^\dagger$$

$$(p(\vec{x}))^\dagger = p(\vec{x})$$

The ask operation  $c \rightarrow A$  has no CLP equivalent.

It is a new **synchronization primitive** between agents.

# CC Computations

Concurrency = communication (shared variables)  
+ synchronization (ask)

Communication channels, i.e. variables, are **transmissible** by agents (like in  $\pi$ -calculus, unlike CCS, CSP, Occam,...)

Communication is additive (a constraint will never be removed), **monotonic accumulation** of information in the store (as in CLP, as in Scott's information systems)

Synchronization makes computation both **data-driven and goal-directed**.

No private communication, all agents sharing a variable will see a constraint posted on that variable,

Not a parallel implementation model.

# CC( $\mathcal{X}$ ) Configurations

Configuration  $(\vec{x}; c; \Gamma)$ : store  $c$  of constraints, multiset  $\Gamma$  of agents, modulo  $\equiv$  the smallest congruence s.t.:

$$\mathcal{X}\text{-equivalence} \quad \frac{c \dashv\vdash_{\mathcal{X}} d}{c \equiv d}$$

$$\alpha\text{-Conversion} \quad \frac{z \notin \text{fv}(A)}{\exists y A \equiv \exists z A[z/y]}$$

$$\text{Parallel} \quad (\vec{x}; c; A \parallel B, \Gamma) \equiv (\vec{x}; c; A, B, \Gamma)$$

$$\text{Hiding} \quad \frac{y \notin \text{fv}(c, \Gamma)}{(\vec{x}; c; \exists y A, \Gamma) \equiv (\vec{x}, y; c; A, \Gamma)} \quad \frac{y \notin \text{fv}(c, \Gamma)}{(\vec{x}, y; c; \Gamma) \equiv (\vec{x}; c; \Gamma)}$$

# CC( $\mathcal{X}$ ) Transitions

Interleaving semantics

$$\text{Procedure call} \quad \frac{(p(\vec{y}) = A) \in \mathcal{D}}{(\vec{x}; c; p(\vec{y}), \Gamma) \longrightarrow (\vec{x}; c; A, \Gamma)}$$

$$\text{Tell} \quad (\vec{x}; c; \text{tell}(d), \Gamma) \longrightarrow (\vec{x}; c \wedge d; \Gamma)$$

$$\text{Ask} \quad \frac{c \vdash_{\mathcal{X}} d[\vec{t}/\vec{y}]}{(\vec{x}; c; \forall \vec{y}(d \rightarrow A), \Gamma) \longrightarrow (\vec{x}; c; A[\vec{t}/\vec{y}], \Gamma)}$$

$$\begin{aligned} \text{Blind choice} & \quad (\vec{x}; c; A + B, \Gamma) \longrightarrow (\vec{x}; c; A, \Gamma) \\ \text{(local/internal)} & \quad (\vec{x}; c; A + B, \Gamma) \longrightarrow (\vec{x}; c; B, \Gamma) \end{aligned}$$

CC( $\mathcal{X}$ ) extra rules

**Guarded choice**  
(global/external)

$$\frac{c \vdash_{\mathcal{X}} c_j}{(\vec{x}; c; \sum_i c_i \rightarrow A_i, \Gamma) \longrightarrow (\vec{x}; c; A_j, \Gamma)}$$

**AskNot**

$$\frac{c \vdash_{\mathcal{X}} \neg d}{(\vec{x}; c; \forall \vec{y}(d \rightarrow A), \Gamma) \longrightarrow (\vec{x}; c; \Gamma)}$$

**Sequentiality**

$$\frac{(\vec{x}; c; \Gamma) \longrightarrow (\vec{x}; d; \Gamma')}{(\vec{x}; c; (\Gamma; \Delta), \Phi) \longrightarrow (\vec{x}; d; (\Gamma'; \Delta), \Phi)}$$

$$(\vec{x}; c; (\emptyset; \Gamma), \Delta) \longrightarrow (\vec{x}; d; \Gamma, \Delta)$$

## Properties of CC Transitions (1)

### Theorem (Monotonicity)

*If  $(\vec{x}; c; \Gamma) \rightarrow (\vec{y}; d; \Delta)$  then  $(\vec{x}; c \wedge e; \Gamma, \Sigma) \rightarrow (\vec{y}; d \wedge e; \Delta, \Sigma)$  for every constraint  $e$  and agents  $\Delta$ .*

### Proof.

*tell and ask are monotonic (monotonic conditions in guards).  $\square$*

### Corollary

*Strong fairness and weak fairness are equivalent.*

## Properties of CC Transitions (2)

A configuration without  $+$  is called **deterministic**.

### Theorem (Confluence)

*For any deterministic configuration  $\kappa$  with deterministic declarations,*

*if  $\kappa \rightarrow \kappa_1$  and  $\kappa \rightarrow \kappa_2$  then  $\kappa_1 \rightarrow \kappa'$  and  $\kappa_2 \rightarrow \kappa'$  for some  $\kappa'$ .*

### Corollary

*Independence of the scheduling of the execution of parallel agents.*



## Properties of CC Transitions (3)

### Theorem (Extensivity)

*If  $(\vec{x}; c; \Gamma) \rightarrow (\vec{y}; d; \Delta)$  then  $\exists \vec{y}d \vdash_{\mathcal{X}} \exists \vec{x}c$ .*

### Proof.

For any constraint  $e$ ,  $c \wedge e \vdash_{\mathcal{X}} c$ .

### Theorem (Restartability)

*If  $(\vec{x}; c; \Gamma) \rightarrow^* (\vec{y}; d; \Delta)$  then  $(\vec{x}; \exists \vec{y}d; \Gamma) \rightarrow^* (\vec{y}; d; \Delta)$ .*

### Proof.

By extensivity and monotonicity.

CC( $\mathcal{X}$ ) Operational Semanticssss

- observing the set of **success stores**,

$$\mathcal{O}_{ss}(\mathcal{D}.A; c) = \{\exists \vec{x}d \in \mathcal{X} \mid (\emptyset; c; A) \longrightarrow^* (\vec{x}; d; \epsilon)\}$$

- observing the set of **terminal stores** (successes and suspensions),

$$\mathcal{O}_{ts}(\mathcal{D}.A; c) = \{\exists \vec{x}d \in \mathcal{X} \mid (\emptyset; c; A) \longrightarrow^* (\vec{x}; d; \Gamma) \dashv\rightarrow\}$$

- observing the set of **accessible stores**,

$$\mathcal{O}_{as}(\mathcal{D}.A; c) = \{\exists \vec{x}d \in \mathcal{X} \mid (\emptyset; c; A) \longrightarrow^* (\vec{x}; d; B)\}$$

- observing the set of **limit stores**?

$$\mathcal{O}_{\infty}(\mathcal{D}.A; c_0) = \{\sqcup? \{\exists \vec{x}_i c_i\}_{i \geq 0} \mid (\emptyset; c_0; A) \longrightarrow (\vec{x}_1; c_1; \Gamma_1) \longrightarrow \dots\}$$

## CC( $\mathcal{H}$ ) 'append' Program(s)

Unidirectional CLP style

Directional CC success store style

Directional CC terminal store style

## CC( $\mathcal{H}$ ) 'append' Program(s)

### Unidirectional CLP style

$$\begin{aligned} \text{append}(A, B, C) = & \text{tell}(A = []) \parallel \text{tell}(C = B) \\ & + \text{tell}(A = [X|L]) \parallel \text{tell}(C = [X|R]) \parallel \text{append}(L, B, R) \end{aligned}$$

### Directional CC success store style

### Directional CC terminal store style

## CC( $\mathcal{H}$ ) 'append' Program(s)

### Unidirectional CLP style

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### Directional CC success store style

### Directional CC terminal store style

CC( $\mathcal{H}$ ) 'append' Program(s)

## Undirectional CLP style

$$\begin{aligned} \text{append}(A, B, C) = & \text{tell}(A = []) \parallel \text{tell}(C = B) \\ & + \text{tell}(A = [X|L]) \parallel \text{tell}(C = [X|R]) \parallel \text{append}(L, B, R) \end{aligned}$$

## Directional CC success store style

$$\begin{aligned} \text{append}(A, B, C) = & (A = [] \rightarrow \text{tell}(C = B)) \\ & + \forall X, L (A = [X|L] \rightarrow \text{tell}(C = [X|R]) \parallel \text{append}(L, B, R)) \end{aligned}$$

## Directional CC terminal store style

CC( $\mathcal{H}$ ) 'append' Program(s)

## Unidirectional CLP style

$$\begin{aligned} \text{append}(A, B, C) = & \text{tell}(A = []) \parallel \text{tell}(C = B) \\ & + \text{tell}(A = [X|L]) \parallel \text{tell}(C = [X|R]) \parallel \text{append}(L, B, R) \end{aligned}$$

## Directional CC success store style

$$\begin{aligned} \text{append}(A, B, C) = & (A = [] \rightarrow \text{tell}(C = B)) \\ & + \forall X, L (A = [X|L] \rightarrow \text{tell}(C = [X|R]) \parallel \text{append}(L, B, R)) \end{aligned}$$

## Directional CC terminal store style

CC( $\mathcal{H}$ ) 'append' Program(s)

## Undirectional CLP style

$$\begin{aligned} \text{append}(A, B, C) = & \text{tell}(A = []) \parallel \text{tell}(C = B) \\ & + \text{tell}(A = [X|L]) \parallel \text{tell}(C = [X|R]) \parallel \text{append}(L, B, R) \end{aligned}$$

## Directional CC success store style

$$\begin{aligned} \text{append}(A, B, C) = & (A = [] \rightarrow \text{tell}(C = B)) \\ & + \forall X, L (A = [X|L] \rightarrow \text{tell}(C = [X|R]) \parallel \text{append}(L, B, R)) \end{aligned}$$

## Directional CC terminal store style

$$\begin{aligned} \text{append}(A, B, C) = & A = [] \rightarrow \text{tell}(C = B) \\ & \parallel \forall X, L (A = [X|L] \rightarrow \text{tell}(C = [X|R]) \parallel \text{append}(L, B, R)) \end{aligned}$$



CC( $\mathcal{H}$ ) 'merge' Program

## Merging streams

$$\begin{aligned} \text{merge}(A, B, C) &= (A = [] \rightarrow \text{tell}(C = B)) \\ &+ (B = [] \rightarrow \text{tell}(C = A)) \\ &+ \forall X, L(A = [X|L] \rightarrow \text{tell}(C = [X|R]) \parallel \text{merge}(L, B, R)) \\ &+ \forall X, L(B = [X|L] \rightarrow \text{tell}(C = [X|R]) \parallel \text{merge}(A, L, R)) \end{aligned}$$

Good for the      observable(s?)

Many-to-one communication:

*client*( $C_1, \dots$ )

...

*client*( $C_n, \dots$ )

*server*( $[C_1, \dots, C_n], \dots$ ) =

$$\sum_{i=1}^n \forall X, L(C_i = [X|L] \rightarrow \dots \parallel \text{server}([C_1, \dots, L, \dots, C_n], \dots))$$

CC( $\mathcal{H}$ ) 'merge' Program

## Merging streams

$$\begin{aligned} \text{merge}(A, B, C) &= (A = [] \rightarrow \text{tell}(C = B)) \\ &+ (B = [] \rightarrow \text{tell}(C = A)) \\ &+ \forall X, L(A = [X|L] \rightarrow \text{tell}(C = [X|R]) \parallel \text{merge}(L, B, R)) \\ &+ \forall X, L(B = [X|L] \rightarrow \text{tell}(C = [X|R]) \parallel \text{merge}(A, L, R)) \end{aligned}$$

Good for the  $\mathcal{O}_{ss}$  observable

Many-to-one communication:

$\text{client}(C1, \dots)$

...

$\text{client}(Cn, \dots)$

$\text{server}([C1, \dots, Cn], \dots) =$

$$\sum_{i=1}^n \forall X, L(Ci = [X|L] \rightarrow \dots \parallel \text{server}([C1, \dots, L, \dots, Cn], \dots))$$

# CC( $\mathcal{FD}$ ) Finite Domain Constraints

Approximating *ask* condition with the Elimination condition

**EL:**  $c \wedge \Gamma \longrightarrow \Gamma$

if  $\mathcal{FD} \models c\sigma$  for every valuation  $\sigma$  of the variables in  $c$  by values of their domain.

$$\text{ask}(X \geq Y + k) = \min(X) \geq \max(Y) + k$$

$$\text{asknot}(X \geq Y + k) = \max(X) < \min(Y) + k$$

$$\text{ask}(X \neq Y) = \max(X) < \min(Y) \vee \min(X) > \max(Y)$$

a better approximation:

$$= (\text{dom}(X) \cap \text{dom}(Y) = \emptyset)$$

# CC( $\mathcal{FD}$ ) Constraints

Basic constraints

$$(X \geq Y + k) = X \text{ in } \min(Y) + k .. \infty \parallel Y \text{ in } 0 .. \max(X) - k$$

Reified constraints

$$(B \Leftrightarrow X = A) = B \text{ in } 0..1 \parallel \\ X = A \rightarrow B = 1 \parallel X \neq A \rightarrow B = 0 \parallel \\ B = 1 \rightarrow X = A \parallel B = 0 \rightarrow X \neq A$$

Higher-order constraints

$$\text{card}(N, L) = L = [] \rightarrow N = 0 \parallel \\ L = [C|S] \rightarrow \\ \exists B, M (B \Leftrightarrow C \parallel N = B + M \parallel \text{card}(M, S))$$

# Andora Principle

“Always execute deterministic computation first”.

Disjunctive scheduling:

deterministic propagation of the disjunctive constraints for which one of the alternatives is dis-entailed:

$$\text{card}(1, [x \geq y + d_y, y \geq x + d_x])$$

before creating choice points:

$$(x \geq y + d_y) + (y \geq x + d_x)$$

# Constructive Disjunction in CC( $\mathcal{FD}$ ) (1)

$$\vee L \quad \frac{c \vdash_{\mathcal{X}} e \quad d \vdash_{\mathcal{X}} e}{c \vee d \vdash_{\mathcal{X}} e}$$

Intuitionistic logic tells us we can *infer the common information* to both branches of a disjunction **without creating choice points!**

$\max(X, Y, Z) = (X > Y \parallel Z = X) + (X \leq Y \parallel Z = Y)$

or

$\max(X, Y, Z) = X > Y \rightarrow Z = X + X \leq Y \rightarrow Z = Y.$

or

$\max(X, Y, Z) = X > Y \rightarrow Z = X \parallel X \leq Y \rightarrow Z = Y.$

better?

# Constructive Disjunction in CC( $\mathcal{FD}$ ) (1)

$$\forall L \quad \frac{c \vdash_{\mathcal{X}} e \quad d \vdash_{\mathcal{X}} e}{c \vee d \vdash_{\mathcal{X}} e}$$

Intuitionistic logic tells us we can *infer the common information* to both branches of a disjunction **without creating choice points!**

$\max(X, Y, Z) = (X > Y \parallel Z = X) + (X \leq Y \parallel Z = Y)$

or

$\max(X, Y, Z) = X > Y \rightarrow Z = X + X \leq Y \rightarrow Z = Y.$

or

$\max(X, Y, Z) = X > Y \rightarrow Z = X \parallel X \leq Y \rightarrow Z = Y.$

better?

$\max(X, Y, Z) = Z \text{ in } \min(X).. \infty \parallel Z \text{ in } \min(Y).. \infty$   
 $\parallel Z \text{ in } \text{dom}(X) \cup \text{dom}(Y)$

## Constructive Disjunction in CC( $\mathcal{FD}$ ) (2)

### Disjunctive precedence constraints

$$\begin{aligned} \text{disjunctive}(T1, D1, T2, D2) = \\ (T1 \geq T2 + D2) + \\ (T2 \geq T1 + D1) \end{aligned}$$

### Using constructive disjunction



# Constructive Disjunction in CC( $\mathcal{FD}$ ) (2)

## Disjunctive precedence constraints

$$\begin{aligned} \text{disjunctive}(T1, D1, T2, D2) = \\ (T1 \geq T2 + D2) + \\ (T2 \geq T1 + D1) \end{aligned}$$

## Using constructive disjunction

$$\begin{aligned} \text{disjunctive}(T1, D1, T2, D2) = \\ T1 \text{ in } (0..max(T2) - D1) \cup (min(T2) + D2..∞) \parallel \\ T2 \text{ in } (0..max(T1) - D2) \cup (min(T1) + D1..∞) \end{aligned}$$

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