Combining Constraint Languages via Abstract Interpretation ICTAI 2019

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#### Context: Project AbSolute (Pelleau and al., 2013)

- Mixed constraint solver on integer and real numbers in OCaml.
- Based on abstract interpretation (especially abstract domains) and constraint programming (CP).
- Abstract domains = partially ordered sets with operators.
- CP  $\cup$  abstract domains  $\cup$  reduced products.
- ► Abstract interpretation \ widenings ∪ backtracking.

Constraint satisfaction problem (CSP)

A CSP is a pair  $\langle d, C \rangle$ , example :

$$\{x \mapsto \{1, 2, 3\}, y \mapsto \{1, 2, 3\}\}, \{x > y, x \neq 2\}$$

A solution is  $\{x \mapsto 3, y \mapsto 1\}$ .

A classic solver in CP:

1: 
$$\operatorname{solve}(\langle d, C \rangle)$$
  
2:  $\langle d', C \rangle \leftarrow \operatorname{propagate}(\langle d, C \rangle)$   
3: if  $d' = \{a\}$  then  
4: return  $\{a\}$   
5: else if  $d' = \{\}$  then  
6: return  $\{\}$   
7: else  
8:  $\langle d_1, \ldots, d_n \rangle \leftarrow \operatorname{branch}(d')$   
9: return  $\bigcup_{i=0}^n \operatorname{solve}(\langle d_i, C \rangle)$   
10: end if

## Classic solver VS solver by abstract interpretation

A solver by abstract interpretation, with Abs an abstract domain::

1: solve( $a \in Abs$ ) 2:  $a \leftarrow closure(a)$ 3: if state(a) = true then 4: return {a} 5: else if state(a) = false then 6: return {} 7: else 8:  $\langle a_1, \ldots, a_n \rangle \leftarrow split(a)$ 9: return  $\bigcup_{i=0}^n solve(a_i)$ 10: end if

Conservative extension: We encapsulate a CSP in an abstract domain.

Global constraints are crucial to efficiently solve CSP but:

- There are a lot (> 400).
- Most of these are very specialized.
- Only a small subset is implemented in mainstream solvers.

We should think about more general methods. We propose to rely on **abstract domains**.

## Global constraints VS abstract domains

- Global constraints: capture a sub-structure + efficient solving algorithm for this structure.
- Abstract domain for CP:
  - Exact representation, or by over-approximation of a constraint language.
  - Partially ordered set equipped with several operations (consistency, entailment, join, ...).
- Various discrete and continuous abstract domains: interval, octagon, polyhedra, etc.
- Combination and transformators over these domains: reduced product, reduced product by reification, partitioning, etc.

# Contributions

- Adapt integer octagon abstract domain to CP.
- Design of a generic reduced product where domains communicate through equivalence constraints.
- We are guided by a scheduling application: Resource-constrained project scheduling problem (RCPSP, RCPSP/max).

Decompose global constraints into abstract domains.



#### Introduction

Scheduling problem RCPSP

► Abstract domains for RCPSP

► Conclusion

NP-complete optimisation problem:

- ▶ *T* is a set of tasks,  $d_i \in \mathbb{N}$  the duration of task *i*.
- P are the precedences among tasks: i ≪ j ∈ P if i must terminate before j starts.
- ▶ *R* is a set of resources where  $k \in R$  has a capacity  $c_k \in \mathbb{N}$ .
- Each task *i* uses a quantity  $r_{k,i}$  of resources *k*.

Goal: find a (minimal) planning of tasks T that satisfies precedences in P without exceeding the capacity of available resources.

### Example with 5 tasks and 2 resources



Resources consumption



Time units

#### Constraints model

- Variables :  $s_i \in \{0..h 1\}$  is the starting time of task *i*.
- Constraints :

$$\forall (i \ll j) \in P, \ s_i + d_i \leq s_j \tag{1}$$

$$\forall j \in [1..n], \ \forall i \in [1..n] \setminus \{j\}, \\ b_{i,j} \Leftrightarrow (s_i \le s_j \land s_j < s_i + d_i)$$

$$(2)$$

$$\forall j \in [1..n], \ r_{k,j} + (\sum_{i \in [1..n] \setminus \{j\}} r_{k,i} * b_{i,j}) \le c_k$$
(3)

- 1. Temporal constraints (eq. 1)
- 2. Resources constraints (eq. 2 and 3): *tasks decomposition* of cumulative.

### Three kinds of constraints

- In green: octagonal constraints treated by octagon abstract domain.
- In red: equivalence constraints treated in a specialized reduced product.
- In blue: interval constraints treated by the CSP abstract domain.

 $\forall (i \ll j) \in P, \ s_i + d_i \leq s_j$ 

$$\forall j \in [1..n], \forall i \in [1..n] \setminus \{j\}, \\ b_{i,j} \Leftrightarrow (s_i \le s_j \land s_j < s_i + d_i) \\ \forall j \in [1..n], r_{k,j} + (\sum_{i \in [1..n] \setminus \{j\}} r_{k,i} * b_{i,j}) \le c_k$$

Equivalence constraints **connect** the CSP and octagon abstract domains.



#### Introduction

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Lattice  $\langle Abs, \leq \rangle$  representable in a machine where:

 $\blacktriangleright$   $\perp$  is the smallest element.

...

- ▶ □ performs the union (*join*) of two elements.
- $[\![.]\!]: \Phi \to Abs$  is a partial function turning a constraint into an element of the abstract domain.
- closure :  $Abs \rightarrow Abs$  propagates the constraints in the abstract domain.
- $\models : Abs \times \Phi \text{ where } a \vDash \varphi \text{ holds if } \gamma(a) \leq \llbracket \varphi \rrbracket^{\natural}.$

# Integer octagon (Miné, 2004)

An integer octagon is defined over a set of variables  $(x_0, \ldots, x_{n-1})$  and constraints:

$$\pm x_i - \pm x_j \leq d$$

where  $d \in \mathbb{Z}$  is a constant.

Complexity of the main operations:

- ▶ join is  $\mathcal{O}(n^2)$ .
- ▶ closure: Floyd-Warshall algorithm in  $\mathcal{O}(n^3)$ , incremental version in  $\mathcal{O}(n^2)$  to add a single constraint (Chawdhary and al., 2018). Normal form equivalent to **path consistency** (Dechter and al., 1991).
- $o \vDash \varphi$  is in constant time when  $\varphi$  is a single octagonal constraint.

### Example of integer octagon

Take the following constraints:

$$egin{array}{ll} x_0 \geq 1 \wedge x_0 \leq 3 & x_1 \geq 1 \wedge x_1 \leq 4 \ x_0 - x_1 \leq 1 & -x_0 + x_1 \leq 1 \end{array}$$

Bound constraints on  $x_0$  and  $x_1$  are represented by the yellow box, and octagonal constraints by the green box.



We can define a direct product over  $\textit{CSP} \times \textit{Oct}$  as follows:

$$(csp, o) \sqcup (csp', o') = (csp \sqcup_{CSP} csp', o \sqcup_{Oct} o')$$
$$\llbracket c \rrbracket = \begin{cases} (\llbracket c \rrbracket_{CSP}, \llbracket c \rrbracket_{Oct}) \\ (\llbracket c \rrbracket_{CSP}, \bot_{Oct}) & \text{if } \llbracket c \rrbracket_{Oct} \text{ is not defined} \\ (\bot_{CSP}, \llbracket c \rrbracket_{Oct}) & \text{if } \llbracket c \rrbracket_{CSP} \text{ is not defined} \\ closure((csp, o)) = (closure(csp), closure(o)) \end{cases}$$

**Issue**: domains do not exchange information.

### Reduced product via equivalence constraints

We consider a reduced product to connect constraints from both domains via equivalence constraints.

▶ Let  $c_1 \Leftrightarrow c_2$  be an equivalence constraint where  $[[c_1]]_{CSP}$  and  $[[c_2]]_{Oct}$  are defined, then we have:

$$\begin{array}{l} \text{prop}_{\Leftrightarrow}(csp, o, c_{1} \Leftrightarrow c_{2}) \triangleq \\ \left\{ \begin{array}{l} csp \vDash_{CSP} c_{1} \implies (csp, o \sqcup \llbracket c_{2} \rrbracket_{Oct}) \\ csp \vDash_{CSP} \neg c_{1} \implies (csp, o \sqcup \llbracket \neg c_{2} \rrbracket_{Oct}) \\ o \vDash_{Oct} c_{2} \implies (csp \sqcup \llbracket c_{1} \rrbracket_{CSP}, o) \\ o \vDash_{Oct} \neg c_{2} \implies (csp \sqcup \llbracket \neg c_{1} \rrbracket_{CSP}, o) \\ (csp, o) \text{ otherwise} \end{array} \right.$$

This propagator is sound: it does not remove solutions.

### Reduced product via equivalence constraints

We improve the closure operator by propagating the set of equivalence constraints R.

Closure operator of the reduced product  $CSP \times Oct$ :

$$closure_R(csp, oct, R) = (\bigsqcup_{r \in R} prop_{\Leftrightarrow}(csp, oct, r), R)$$

Let e be an element of the reduced product, then the closure operator can be applied to a fixed point closure(e) = e.

**Result**: A generic reduced product to combine abstract domains with disjoint set of variables.



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### Benchmarks

- In brief: We are bettered by state of the art methods (e.g. CHUFFED with *lazy clause*).
- In comparison to GeCode/cumulative:

Problem : rcpsp-max - Instance set : sm\_j20 - Number of instances : 270

inter exter exter only gecode-6.1.0 with rcpsp-cumulative-min\_lb.csv only absolute-2d33cd7 with Octagon-Min\_max\_LB.csv



### Related work

Satisfiability modulo theories (SMT)

- SMT theories and abstract domains very close but on different underlying theories (logic VS posets).
- Implementation of Nelson-Oppen usually not formalized and mysterious.
- Abstract Conflict Driven Learning (D'Silva et al., 2013).
  - Very nice theoretical framework to integrate solving and abstract interpretation.
  - Still a big gap between theory and practice.

We aim to reduce the gap between practice and theory.

# Conclusion

#### 1. New structures:

- Investigate integer octagon abstract domains for CP.
- A new combination: reduced product by equivalence constraints.
   ⇒ It allows us to use reified constraints in abstract domains.
- 2. A case study: RCPSP.

**Bonus**: We can handle **continuous temporal constraint** with discrete resources.

3. Automated benchmarking framework (including Chuffed, GeCode, AbSolute).

# O github.com/ptal/AbSolute/tree/ictai2019

# Thank you!

