Modular Constraint Solver Cooperation via Abstract Interpretation ICLP 2020

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Introduction

Many research communities centered around solving techniques and constraint languages:

- SAT solving: propositional formulas,
- Linear programming: linear relations either on real numbers, integers, or both (mixed),
- Constraint programming: Boolean and arithmetic constraints with specialized predicates (global constraints),
- Answer set programming: Horn clauses w/o functions (initially),
 ...

Even larger if we consider heuristics approaches such as genetic algorithms, evolutionary algorithms or local search.

The challenge

- Each field has developed its own theory and terminology.
- Pro: Very specialized and efficient on their constraint languages.
- Drawback: Hard to transfer knowledge from one field to another.

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The overarching project

Take a step back, and try to find a unified theory.

Abstract interpretation is a framework to statically analyse programs and catch bugs (*Cousot and Cousot, 1977*).

Interesting features for constraint reasoning

- Mathematical background on lattice theory.
- Abstract domains are lattices with operators encapsulating a constraint language.
- Product of domains to combine several abstract domains, thus constraint solving techniques.
- Under-approximation and over-approximation to characterize the solutions of an abstract element (soundness and completeness).

AbSolute: A constraint solver written in OCaml to experiment our ideas.

- 2013: Constraint solver with linear programming, constraint programming, temporal reasoning (*Pelleau and al., 2013*).
 - Mostly over continuous domains, using over-approximations.
 - Cartesian product among abstract domains.
- 2019: Under-approximation and discrete constraint solving, with logical combination of abstract domains (*Talbot and al., 2019*).

Focus on domain transformers: abstract domains parametrized by other abstract domains.

Contributions

- Two domain transformers to combine abstract domains sharing variables.
 - 1. Interval propagators completion: Arithmetic constraints over product of domains.
 - 2. Delayed product: Exchange of over-approximations among abstract domains.
- Shared product to combine domain transformers.

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Abstract interpretation in a nutshell



An example



Over-approximation



Under-approximation



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

- \blacktriangleright \leq is the order, where $a \leq b$ if b "contains more information than" a,
- \blacktriangleright \perp is the smallest element, \sqcup the join, . . .
- $\blacktriangleright \ \llbracket . \rrbracket^{\sharp} : \Phi \to A \text{ and } \gamma : A \to C^{\flat},$
- closure : $A \rightarrow A$ to refine an abstract element,
- ▶ *split* : $A \rightarrow \mathcal{P}(A)$ to divide an element into sub-elements,
- state : A → {true, false, unknown} to retreive the "solving state" of an element.

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

```
1: solve(a \in A)
 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 \text{split}(a)
 9: return \bigcup_{i=0}^{n} \operatorname{solve}(a_i)
10: end if
```

We call $solve(\llbracket \varphi \rrbracket^{\sharp})$ to obtain the solutions of the formula φ .

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Direct product: combination of abstract domains

Consider the formula $\varphi \triangleq x > 4 \land x < 7 \land y + z \le 4$.

> $x > 4 \land x < 7$ can be treated in the box abstract domain *Box*,

▶ $y + z \le 4$ can be treated in the octagon abstract domain *Octagon*.

Solution: Rely on the direct product $Box \times Octagon$.

Direct product: combination of abstract domains

Consider the formula $\varphi \triangleq x > 4 \land x < 7 \land y + z \le 4$.

 $x > 4 \land x < 7$ can be treated in the box abstract domain *Box*,

▶ $y + z \le 4$ can be treated in the octagon abstract domain *Octagon*. **Solution**: Rely on the direct product *Box* × *Octagon*.

Direct product

 $\langle A_1 \times \ldots \times A_n, \leq \rangle$ is an abstract domain where each operator is defined coordinatewise:

(a₁,..., a_n) ≤ (b₁,..., b_n) ⇔ ∧_{1≤i≤n} a_i ≤_i b_i

$$\gamma((a_1,..., a_n)) \triangleq \bigcup_{1 ≤ i ≤ n} \gamma_i(a_i)$$
closure((a₁,..., a_n)) ≜ (closure₁(a₁),..., closure_n(a_n))
...

Issue: domains do not exchange information.

Consider the constraint $\varphi \triangleq x > 1 \land x + y + z \le 5 \land y - z \le 3$.

- x > 1 can be interpreted in boxes,
- ▶ $y z \leq 3$ in octagons,
- but $x + y + z \le 5$ is too general for any of these two...
- …and it shares its variables with the other two.

Solution: Use the notion of *propagator functions* to connect variables between abstract domains.

Example: Propagator $x \ge y$

We assume a projection function $project : A \times Vars \rightarrow I$, $project(a, x) = [x_{\ell}..x_u]$ and $project(a, y) = [y_{\ell}..y_u]$:

$$\llbracket x \geq y
rbracket = \lambda a.a \sqcup_{\mathcal{A}} \llbracket x \geq y_{\ell}
rbracket_{\mathcal{A}} \sqcup_{\mathcal{A}} \llbracket y \leq x_u
rbracket_{\mathcal{A}}$$

- IPC(A) = A × P(Prop) is a domain transformer equipping A with propagators,
- We can rely on $IPC(Box \times Octagon)$ with a propagator for $x + y + z \le 5$,
- The bound constraints will automatically be exchanged between both domains thanks to the propagator.

IPC exchanges bound constraints, can we do better?

- $\varphi \triangleq x > 1 \land x + y + z \le 5 \land y z \le 3.$
- **Observation**: When x is instantiated in $x + y + z \le 5$, we can transfer the constraint in octagons.
- ▶ We have the *delayed product* $DP(A_1, A_2)$ to transfer instantiated constraints from A_1 into a more specialized abstract domain A_2 .
- For instance, consider the abstract domain DP(IPC(Box × Octagon), Octagon), whenever x = 3, we can transfer 3 + y + z ≤ 5 into the octagon.

Delayed product (improved closure)

Even better?

- $\varphi \triangleq x > 1 \land x + y + z \le 5 \land y z \le 3.$
- **Observation**: We can transfer *over-approximations* of $x + y + z \le 5$ in octagons.
- For instance, if x = [1..3], we can transfer 1 + y + z ≤ 5 ⇔ y + z ≤ 4 into the octagon.
- A solution of y + z ≤ 4 will also be a solution of x + y + z ≤ 5, since x must be at least equal to 1.
- Formally: $\gamma(a \sqcup [x + y + z \le 5]]^{\sharp}) \subseteq \gamma(a \sqcup [y + z \le 4]]^{\sharp}).$

- Domain transformers combine abstract domain.
- How to combine domain transformers? Especially when they share sub-domains.

Solution: Shared product

- A "top-level" product combining domain transformers and abstract domains.
- ▶ Merge the shared sub-domains in domain transformers using the join ⊔.

Application

We experimented on the flexible job-shop scheduling problem.

- Temporal constraints of the form $x + y \le d$ (with 3 variables).
- We can treat most of the constraints in IPC(Box × Octagon).
- Over-approximations can be sent in octagons for better efficiency.

Results

- Competitive w.r.t. state of the art (Chuffed) on set of instances with few machines.
- Our goal is not (yet) to beat benchmarks, but to prove the feasibility of our approach.

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Related work

Satisfiability modulo theories (SMT)

- Focus on logical properties, abstract domains focus more on semantics and modularity.
- Nelson-Oppen is a fixed cooperation scheme, we can run several cooperation schemes concurrently.
- 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.
- *TOY* (*Estévez-Martín et al., 2009*): notion of bridges among variables, subsumed by *IPC* in our framework.

We aim to reduce the gap between practice and theory.

Conclusion

- Constraint solver = abstract domain.
- Cooperation scheme = domain transformer.
- ▶ We show two cooperation schemes (*IPC* and *DP*).
- The shared product allows us to use several cooperation schemes concurrently and in a modular way.

O github.com/ptal/AbSolute/tree/iclp2020