## Casper documentation

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## 1 Introduction

Casper is a modeling language for constraint programming with declarative search strategy specification, in the line of the Rules2CP language [5]. The Casper compiler transforms Casper models into GNU or Sicstus Prolog programs with FD constraints [3, 2] or Java programs using either the Choco constraint solver [9] or the MiniSAT solver. Search strategies are specified by, on the first hand, unordered decision trees expressed with Boolean expressions, and, on the other hand, an ordering criterion on Boolean trees. This specification is transformed into a conjunction of reified constraints and a labeling ordering on the associated Boolean variables. Section 2 specifies the user-level language, consisting in a typed functional kernel for expressing Boolean expressions on arithmetic constraints. Section 3 defines the transformation from Boolean expressions into reified constraints and labeling ordering and illustrates this transformation for some usual search strategies. Section 4 gives the most relevant parts of the compilation chain from the Casper language to the targeted solvers.

## 2 User-level language

### 2.1 Example: $n$-queens in Casper

The following program models the $n$-queens problem in Casper.

```
let n := parameter('n') in
let queens := [{row: i, column: _} | i in 0 .. n - 1] in
let not_attack(q, q') := (
    q.column != q'.column /\
    q.column + q.row != q'.column + q'.row /\
    q.column - q.row != q'.column - q'.row) in
solve({
    constraints:
        (forall q in queens, 0 <= q.column < n) /\
            (forall q in queens,
                forall q' in queens,
                    q.row != q'.row => not_attack(q, q')),
    search:
            forall q in queens,
                q.column <- inf(q.column) .. sup(q.column)})
```

This example shows the two distinctive traits of Casper with respect to other constraint modelling languages such as Zinc [10] or Essence [6]. The first trait is that the parametricity of

```
\mathcal{G}=|\langlevariable\rangle | \langleinteger \rangle | \langlestring\rangle | _ | (\mathcal{G)}
    |\mathcal{G}\bowtie\mathcal{G}}\bowtie\in{{+,-,*,/
    |\mathcal{G}\bowtie\mathcal{G}\bowtie\in{=,!=,<,<=,>,>=}
    |inf(\mathcal{G})|\operatorname{sup}(\mathcal{G})\quad\mathrm{ indexicals}
    not(G)
    \mathcal{G}\bowtie\mathcal{G}}\bowtie\in{/\,\/,=>
    |\mathcal{G},\ldots,\mathcal{G}] list constructor
    |\mathcal{G | \langlevariable\rangle in \mathcal{G}}]
    G..G
    |G[G]
    |forall <variable\rangle in \mathcal{G,G}
    | exists \langlevariable\rangle in \mathcal{G},\mathcal{G}
    | sum \langlevariable\rangle in \mathcal{G, G}
    | product \langlevariable\rangle in \mathcal{G, G}
    |{\langleield\rangle:\mathcal{G},\ldots,\langlefield\rangle:\mathcal{G}} record constructor
    G.\{ield\rangle record projection
    l let \langlevariable\rangle :=\mathcal{G}}\mathrm{ in }\mathcal{G}\quadlocal definition of value
    | let \langlevariable\rangle(\langlevariable\rangle, ..., \langlevariable\rangle) := \mathcal{G}}\mathrm{ in }\mathcal{G}\mathrm{ local definition of functions
    | \langlevariable\rangle(\langlevariable\rangle, ..., \langlevariable\rangle) function application
    | \langlevariable\rangle<-\mathcal{G}}\mathrm{ finite-domain labeling
    | once(G) extra-logical cut
    | solve(G) solve directive
    | parameter(\langlename\rangle) access to model parameters
    | import(\langlefilename\rangle) basic module system
```

Table 1: Casper grammar
the model (the variable $n$ taken as a parameter) is preserved in the generated intermediary code and is handled by the targetted back-end. The second trait is that the model specifies the search strategy: the example shows a simple enumerating labeling from lower-bound to upper-bound for each variable.

### 2.2 Casper grammar

The grammar of Casper expressions is given in table 1. A simple type-system over these expressions is defined in section 2.3. A Casper file may contain any Casper expression and a Casper program is a file that contain an expression of type goal. The program can refer to values defined in other files with the import directive import (' $\langle$ filename $\rangle$ ').

An expression of type goal is constructed with the solve directive that takes a record in argument: solve(\{ constraints: $c$, search: $s$ \}). This record specifies the two components that define a constraint solving problem: the constraints $c$ and the search strategy $s$. The expressions $c$ and $s$ are both of type $\operatorname{var}$ (bool), the type of (possibly uninstantiated) Boolean expressions. In the constraints expression $c$, Boolean connectives are interpreted as constraints. In the search strategy $s$, Boolean conjunction $八$ is interpreted as a sequence and Boolean dis-
junction $\backslash /$ is interpreted as a choice-point (and Boolean implication is not allowed).
The construction $x<-$ list is used in different contexts: basically, $x<-\left[e_{1}, \ldots, e_{n}\right]$ is interpreted as $x=e_{1} \backslash / \ldots \backslash / x=e_{n}$. Consequently, in constraint expressions, $x<-l \ldots$ $u$ gives the domain $\{l, \ldots, u\}$ to the variable $x$. In the search strategy, indexicals for a finitedomain variable $x$ can be accessed with $\inf (x)$ and $\sup (x)$ : we use them in conjunction with the $x<-$ list expression in order to express basic enumeration strategies (e.g., $x<-\inf (x)$ .. $\sup (x)$ for a labeling from the lower bound to the upper bound). The Casper small-step semantics is defined in section 2.4 and the expressiveness of the search strategy language is discussed in section 3 .

The import directive import(' $\langle$ filename $\rangle$ ') used in conjunction with files defining record expressions provides Casper with a simple module system. This module system is illustrated by the small standard library that is distributed with Casper: e.g., import('optimization.csp') returns a record containing functions for standard optimization procedures that transform expressions of type goal that solves CSP satisfiability into expression of type goal that implement optimizers (typically, import('optimization.csp').branch_and_bound). Similarly, import('packing.csp') returns a record defining functions for building constraints for packing problems such as import('packing.csp').non_overlap.

### 2.3 Type system

Casper types have the following forms.

- The basic types of Casper are int, bool, string and goal: bool is a subtype of int for the values 0 and 1 .
- The type constructor $\operatorname{var}(\tau)$ denotes the type of finite-domain variables, where $\tau$ is either int or bool. var (bool) is a subtype of var(int), bool is a subtype of $\operatorname{var}$ (bool) and int is a subtype of $\operatorname{var}$ (int). goal is a subtype of $\operatorname{var}$ (bool).
- The type constructor $[\tau]$ denotes the type of lists whose elements are of type $\tau$ : all lists in Casper are homogeneous. If $\sigma$ is a subtype of $\tau$, then $[\sigma]$ is a subtype of $[\tau]$.
- The type constructor $\left\{\left\langle\right.\right.$ field $\left._{1}\right\rangle: \tau_{1}, \ldots,\left\langle\right.$ field $\left.\left._{n}\right\rangle: \tau_{n}\right\}$ denotes the type of records that contain the pair-wise distinct fields $\left\langle\right.$ field $\left._{1}\right\rangle, \ldots\left\langle\right.$ field $\left._{n}\right\rangle$ such that the value of $\left\langle\right.$ field $\left._{i}\right\rangle$ has the type $\tau_{i}$. A record type $\sigma$ is a subtype of a record type of $\tau$ if all the fields in $\tau$ occur in $\sigma$ and if $\left\langle\right.$ field $\left.d_{i}\right\rangle$ has type $\tau_{i}$ in $\tau$, then $\left\langle\right.$ field $\left.d_{i}\right\rangle$ has type $\sigma_{i}$ in $\sigma$ and $\sigma_{i}$ is a subtype of $\tau_{i}$.
- The type constructor $\left(\tau_{1}, \ldots, \tau_{n}\right)->\tau$ denotes the type of functions with $n$ arguments of types $\tau_{1}, \ldots, \tau_{n}$ respectively and whose produce a value of type $\tau$.

The typing rules are given in table 2. It is worth noticing that the type system is monomorphic: for each value (function argument or let-binding), if there is no constraint that forces another type, the type var (int) is chosen by default. Moreover, let-binding allows the definition of recursive values.

Lists of finite-domain variables are allowed (with the type [var(int)] or [var(bool)]). However, the length of the lists is always fixed. In particular, the construction $l . . u$ is only allowed when $l$ and $u$ have the type int (and not $\operatorname{var}$ (int)).

### 2.4 Semantics

The semantics of Casper follows a call-by-value evaluation strategy. All the constructions of fully instantied values (i.e., all the types except $\operatorname{var}(i n t)$ and $\operatorname{var}(\mathrm{bool})$ ) follow the usual

$$
\begin{aligned}
& \frac{b \in\{0,1\}}{\Gamma \vdash b: \text { bool }} \quad \frac{n \geq 2}{\Gamma \vdash n: \text { int }} \quad \overline{\Gamma \vdash\langle\text { string }\rangle: \text { string }} \quad \overline{\Gamma \vdash \_: \operatorname{var(int)}} \quad \overline{\Gamma, x: \tau \vdash x: \tau} \\
& \frac{\Gamma \vdash e: \sigma \quad \sigma \text { subtype of } \tau}{\Gamma \vdash e: \tau} \quad \frac{\Gamma \vdash e_{1}: \text { int } \quad \Gamma \vdash e_{2}: \text { int } \bowtie \in\{+,-, *, /\}}{\Gamma \vdash e_{1} \bowtie e_{2}: \text { int }} \\
& \frac{\Gamma \vdash e_{1}: \operatorname{var}(\text { int }) \quad \Gamma \vdash e_{2}: \operatorname{var}(\text { int })}{\Gamma \vdash \in\{+,-, *, /\}} \\
& \frac{\Gamma \vdash e_{1}: \text { int } \quad \Gamma \vdash e_{2}: \text { int } \quad \bowtie \in\{=,!=,<,<=,>,>=\}}{\Gamma \vdash e_{1} \bowtie e_{2}: \text { bool }} \\
& \frac{\Gamma \vdash e_{1}: \operatorname{var}(\text { int }) \quad \Gamma \vdash e_{2}: \operatorname{var}(\text { int }) \quad \bowtie \in\{=,!=,\langle,\langle=,>,>=\}}{\Gamma \vdash e_{1} \bowtie e_{2}: \operatorname{var}(\text { bool })} \\
& \frac{\Gamma \vdash e: \operatorname{var}(\mathrm{bool})}{\Gamma \vdash \inf (e): \operatorname{bool}} \quad \frac{\Gamma \vdash e: \operatorname{var}(\mathrm{int})}{\Gamma \vdash \inf (e): \operatorname{int}} \quad \frac{\Gamma \vdash e: \operatorname{var}(\mathrm{bool})}{\Gamma \vdash \sup (e): \operatorname{bool}} \quad \frac{\Gamma \vdash e: \operatorname{var}(\mathrm{int})}{\Gamma \vdash \sup (e): \mathrm{int}} \\
& \frac{\Gamma \vdash e: \text { bool }}{\Gamma \vdash \operatorname{not}(e): \text { bool }} \quad \frac{\Gamma \vdash e: \operatorname{var(bool)}}{\Gamma \vdash \operatorname{not}(e): \operatorname{var}(\mathrm{bool})} \quad \frac{\Gamma \vdash e_{1}: \text { bool } \quad \Gamma \vdash e_{2}: \text { bool } \bowtie \in\{\Lambda, \backslash /,=>\}}{\Gamma \vdash e_{1} \bowtie e_{2}: \text { bool }} \\
& \frac{\Gamma \vdash e_{1}: \operatorname{var}(\mathrm{bool}) \quad \Gamma \vdash e_{2}: \operatorname{var}(\mathrm{bool}) \quad \bowtie \in\{八, \backslash /,=>\}}{\Gamma \vdash e_{1} \bowtie e_{2}: \operatorname{var}(\mathrm{bool})} \\
& \frac{\Gamma \vdash e_{1}: \tau \quad \ldots \quad \Gamma \vdash e_{n}: \tau}{\Gamma \vdash\left[e_{1}, \ldots, e_{n}\right]:[\tau]} \frac{\Gamma \vdash e_{2}: \sigma \quad \Gamma, x: \sigma \vdash e_{2}: \tau}{\Gamma \vdash\left[e_{1} \mid x \text { in } e_{2}\right]:[\tau]} \frac{\Gamma \vdash l: \text { int } \quad \Gamma \vdash u: \text { int }}{\Gamma \vdash l \ldots u:[\text { int }]} \frac{\Gamma \vdash l:[\tau] \quad \Gamma \vdash i: \text { int }}{\Gamma \vdash l[i]: \tau} \\
& \frac{\Gamma \vdash e_{1}:[\tau] \quad \Gamma, x: \tau \vdash e_{2}: \text { bool }}{\Gamma \vdash \text { forall } x \text { in } e_{1}, e_{2}: \text { bool }} \quad \frac{\Gamma \vdash e_{1}:[\tau] \quad \Gamma, x: \tau \vdash e_{2}: \operatorname{var}(\text { bool })}{\Gamma \vdash \text { forall } x \text { in } e_{1}, e_{2}: \operatorname{var}(\text { bool })} \\
& \frac{\Gamma \vdash e_{1}:[\tau] \quad \Gamma, x: \tau \vdash e_{2}: \text { bool }}{\Gamma \vdash \text { exists } x \text { in } e_{1}, e_{2}: \text { bool }} \quad \frac{\Gamma \vdash e_{1}:[\tau] \quad \Gamma, x: \tau \vdash e_{2}: \operatorname{var(bool)}}{\Gamma \vdash \text { exists } x \text { in } e_{1}, e_{2}: \operatorname{var}(\text { bool })} \\
& \frac{\Gamma \vdash e_{1}:[\tau] \quad \Gamma, x: \tau \vdash e_{2}: \text { int }}{\Gamma \vdash \operatorname{sum} x \text { in } e_{1}, e_{2}: \text { int }} \quad \frac{\Gamma \vdash e_{1}:[\tau] \quad \Gamma, x: \tau \vdash e_{2}: \operatorname{var(int)}}{\Gamma \vdash \operatorname{sum} x \text { in } e_{1}, e_{2}: \operatorname{var}(\text { int })} \\
& \frac{\Gamma \vdash e_{1}:[\tau] \quad \Gamma, x: \tau \vdash e_{2}: \text { int }}{\Gamma \vdash \text { product } x \text { in } e_{1}, e_{2}: \text { int }} \quad \frac{\Gamma \vdash e_{1}:[\tau] \quad \Gamma, x: \tau \vdash e_{2}: \operatorname{var}(\text { int })}{\Gamma \vdash \operatorname{product} x \text { in } e_{1}, e_{2}: \operatorname{var}(\text { int })} \\
& \frac{\Gamma \vdash e_{1}: \tau_{1} \ldots \ldots \quad \Gamma \vdash e_{n}: \tau_{n}}{\Gamma \vdash\left\{l_{1}: e_{1}, \ldots, l_{n}: e_{n}\right\}:\left\{l_{1}: \tau_{1}, \ldots, l_{n}: \tau_{n}\right\}} \quad \frac{\Gamma \vdash e:\{l: \tau\}}{\Gamma \vdash e . l: \tau} \quad \frac{\Gamma, x: \sigma \vdash e_{1}: \sigma \quad \Gamma, x: \sigma \vdash e_{2}: \tau}{\Gamma \vdash \text { let } x:=e_{1} \text { in } e_{2}: \tau} \\
& \frac{\Gamma, x_{1}: \sigma_{1}, \ldots, x_{n}: \sigma_{n}, f:\left(\sigma_{1}, \ldots, \sigma_{n}\right)->\sigma \vdash e_{1}: \sigma \quad \Gamma, f:\left(\sigma_{1}, \ldots, \sigma_{n}\right)->\sigma \vdash e_{2}: \tau}{\Gamma \vdash \operatorname{let} f\left(x_{1}, \ldots, x_{n}\right):=e_{1} \text { in } e_{2}: \tau} \\
& \frac{\Gamma \vdash f:\left(\sigma_{1}, \ldots, \sigma_{n}\right)->\sigma \quad \Gamma \vdash e_{1}: \sigma_{1} \quad \ldots \quad e_{n}: \sigma_{n}}{\Gamma \vdash f\left(e_{1}, \ldots, e_{n}\right): \sigma} \quad \frac{\Gamma \vdash x: \operatorname{var(int)} \quad \Gamma \vdash l:[\mathrm{int}]}{\Gamma \vdash x<-l: \operatorname{var}(\mathrm{bool})} \\
& \underline{\Gamma \vdash e:\{c o n s t r a i n t s: \operatorname{var}(\text { bool }), \text { search: } \operatorname{var}(\text { bool })\}} \\
& \Gamma \vdash \operatorname{solve}(e): \text { goal }
\end{aligned}
$$

Table 2: Typing rules
functional semantics. Partially instantiated expressions of type var (int) and var (bool) are handled as abstract terms represented as expression trees. There are two back-end specific constructions. The first construction is the handling of parameters: for the implemented backend, the parameters are commonly read from the command-line or, if the command-line does not specify them, then they are asked interactively in the terminal. The second construction is the handling of solve ( $e$ ), which expects that $e$ evaluates into a record with two fields: constraints and search. Both fields should be Boolean expressions of type var(bool). The expression in the field constraints is added in the constraint store of the underlying constraint solver. The expression in the field search is read as a $\wedge / \vee$ decision-tree of elementary constraints. This tree is evaluated depth-first and from left to right: $\backslash$ denotes conjunctions and $\backslash /$ denotes disjunctions. Common search strategies that are expressible in this framework are given in section 3.

## 3 Search strategies

Besides the basic enumeration labeling expressible by conjunctions of $x<-l$ and indexicals, most search strategies are expressible in terms of $\wedge / \vee$ decision-trees.

The following function gives the definition of dichotomic (or bissection) search, where the size of the domain of a variable is reduced by two, recursively.

```
let dichotomy(x) :=
    inf(x) = sup(x) \/
    inf(x)< sup(x) /\ (
        let m := (inf(x) + sup(x)) / 2 in
        (x <= m \/ x > m) /\ dichotomy(x)) in ...
```

It is easy to notice that $\sup (x)-\inf (x)$ strictly decreases at each recursive call: a (rudimentary) arithmetic analysis detects such specific cases and rejects recursions for which a simple terminaison argument cannot be found.

The following function gives the definition of interval splitting, where the size of the domain of a variable is reduced by a fixed factor.

```
let interval_splitting(x, k) :=
    let s := (sup(x) - inf(x)) / k in
    exists i in 0 .. k,
        (x >= i * s /\ x < i * (s + 1)) in ...
```

These definitions are provided in the module enumeration.csp which contains a record, the fields of which contain these functions.

Having goal as a subtype of constraints var (bool) allows decision-trees to express optimization procedures by iterating solver phases. However, there is a need to introduce mutable variables persistent over search points to remember values found in previous search phases: this is done by adding the ref() construction creating a new mutable variable and a binary := goal for assignment.

The following functions are provided in the module optimization.csp. Branch-and-bound is the iteration of a goal for finding a better solution with respect to an objective variable, or proving the optimality, i.e., that there is no better solution.

```
let branch_and_bound(goal, x) :=
    let best := ref(inf(x)) in
    solve(goal) /\ best := inf(x)
    \/ x < best /\ branch_and_bound(goal, x) in ...
```

The termination is guaranteed since $\sup (x)$ strictly decreases $($ since $\inf (x) \leq \sup (x))$. It is worth noticing that if solve (goal) fails, then we have best $=\inf (x)$ and $\mathrm{x}<$ best immediately fails too. Another optimization strategy is on the contrary to try to assign the objective variable to its lowest value first, and then augments the value until a solution is found.

```
let minimize_increasing(goal, x):
    let minimize :=
        let store = ref(inf(x)) in
        (
            x = inf(x) /\
            post_constraint
        )
        \/
        objective > store /\ minimize in ...
```


## 4 Compilation targeting constraint solvers or SAT solvers

There are three implemented back-end for Casper: Prolog with finite-domain variables (with GNU-Prolog [3] and Sicstus [2] variants), Java with the Choco solver [9] and Java with MiniSAT [4].

Besides the arithmetic constraints included in the syntax, solver-specific constraints can be used such as the geost global constraint [1] in Sicstus or Choco for modelling placement problems (packing.csp module).

Lists are implemented as lists in Prolog and as arrays in Java.
In Java, records are compiled into interfaces, with methods for functional fields and getters for other fields. For example, the interface for a Queen, section 2.1, is the following.

```
interface Queen {
    public IntegerVariable getColumn();
    public IntegerVariable getRow();
}
```

The MiniSAT back-end handles only models whose all variables are var (bool) (i.e., models that do not contain $\operatorname{var}$ (int)). The following Boolean example enumerates all the siphons [8] in an SBML model [7].

```
let filename := parameter('filename') in
let graph := import('sbml.csp').load(filename) in
let siphon := [ _ | i in 0 .. length(graph.places) - 1 ] in
let enumeration := import(enumeration.csp') in
(forall i in 0 .. length(graph.places) - 1,
    0 <= siphon[i] <= 1) /\
(forall i in 0 .. length(graph.places) - 1,
    siphon[i] =>
        (forall j in graph.places[i].predecessors,
            forall k in graph.transitions[j].predecessors, siphon[k])) /\
enumeration.enumerate(
    solver({ search: (forall i in 0 .. length(graph.places) - 1, siphon[i] <- 0 .. 1)}))
```

Compilation to MiniSAT is implemented as a two-step compilation, transforming first the model in a Casper model in conjunctive normal form.

## References

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