



Executable Logic Workshop @ UDC

September 14, 2018

J. Arias^{1,2} **M. Carro**^{1,2} **E. Salazar**³ **K. Marple**³ **G. Gupta**³

¹IMDEA Software Institute, ²Universidad Politécnica de Madrid, ³University of Texas at Dallas



Last Activities

- Tabling: advantages of bottom-up computation using top-down execution.
 - Termination, performance.
 - Well known, old (but still challenging)

Last Activities

- Tabling: advantages of bottom-up computation using top-down execution.
 - Termination, performance.
 - Well known, old (but still challenging)
- Add constraints: enhance expressiveness, termination properties, also speed.
 - First implementation with *full* call / answer entailment checks.
 - Modular: constraint solvers as plug-ins.
 - Improved termination results w.r.t. [Toman 1997].
 - PPDP'16, TPLP'1? (submitted).

Last Activities

- Tabling: advantages of bottom-up computation using top-down execution.
 - Termination, performance.
 - Well known, old (but still challenging)
- Add constraints: enhance expressiveness, termination properties, also speed.
 - First implementation with *full* call / answer entailment checks.
 - Modular: constraint solvers as plug-ins.
 - Improved termination results w.r.t. [Toman 1997].
 - PPDP'16, TPLP'1? (submitted).
- Top-down execution of ASP with constraints — evolution of s(ASP).
 - Can execute non-grounded CASP programs.
 - Constraint system with arbitrary (e.g., unbound) variable domains.
 - Partial models: relevance.
 - Almost ASP semantics: *unsafe* goals allowed.
 - ICLP'18.

Tabling

- Solve issues with loops in SLD resolution:

```

1   p(b) :- p(X) .
2   p(a) .
3
4   ?- p(A) .
  
```

- Variant calls *suspend*:
 - Branch freezes.
 - Execution switches to another clause.
 - Possible results feed and resume suspended calls. Execution continues.
- Termination for programs with bounded-depth property.
- Involved implementation!

Tabling + Constraints

- Add constraints: same, plus entailment instead of variance.

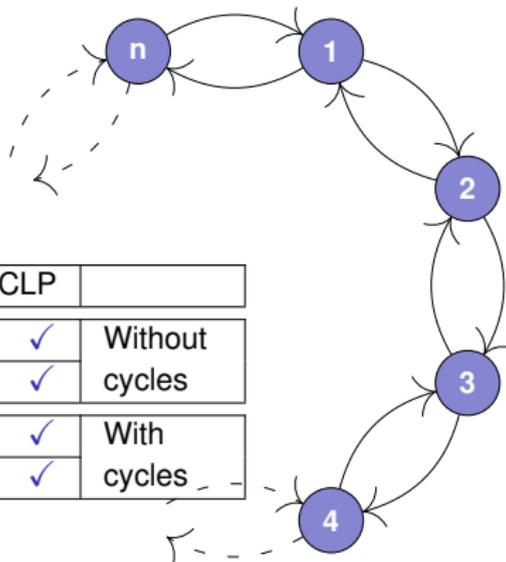
```

1      p(X) :- X #> Y, p(Y) .
2      p(0) .
3
4      ?- p(A) .
  
```

- More particular calls suspend.
 - $p(Y) \{Y < X\}$ more particular than $p(X)$.
 - Suspension, resumption driven by *constraint entailment*.
- Answers checked for entailment: only more general answers kept.
- Less resumptions: speedup.
- Termination guaranteed for compact constraint domains ([Toman 1997])
- Also, for programs that generate compact subset of constraint domain ([Arias & Carro]).
- **Very** involved implementation.

Termination Comparison

Example: Find nodes in a weighted graph within a distance K from each other (using comparable –very similar– programs).



| | Prolog | CLP | Tabling | TCLP | |
|-----------------|--------|-----|---------|------|----------------|
| Left recursion | x | x | ✓ | ✓ | Without cycles |
| Right recursion | ✓ | ✓ | ✓ | ✓ | |
| Left recursion | x | x | x | ✓ | With cycles |
| Right recursion | x | ✓ | x | ✓ | |

Design: Flexibility

- Constraint solver implementations pluggable.
- In general, amounts to writing an interface file to access projection and constraint store extraction.
- Validated with several cases.

CLP(D_{\leq}) Connection of existing constraint solver for difference constraints, written in C.

CLP(Q) and CLP(R) Constraint solvers for linear equations over rationals (CLP(Q)) and over reals (CLP(R)) ([Holzbaur 1995]).

CLP(Lat) New constraint solver over finite lattices.

Performance evaluation (Time)

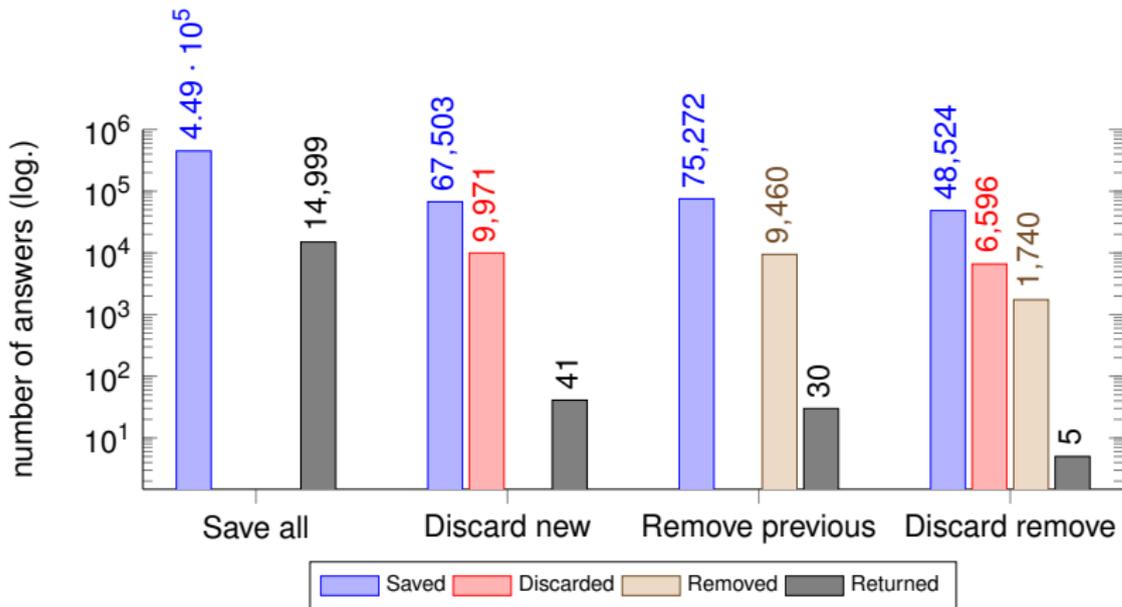
Different answer management strategies

| | CLP(D_{\leq}) | Modular TCLP(D_{\leq}) |
|----------------|-------------------|-------------------------------|
| truckload(300) | 40452 | 7268 |
| truckload(200) | 4179 | 2239 |
| truckload(100) | 145 | 259 |

| | Save all | Discard new answer | Remove previous | Discard and remove |
|----------------|----------|-----------------------|--------------------|-----------------------|
| truckload(300) | 742039 | 7806 | 7780 | 7268 |
| truckload(200) | 11785 | 2314 | 2354 | 2239 |
| truckload(100) | 300 | 263 | 263 | 259 |
| step_bound(30) | – | 8450 | – | 1469 |
| step_bound(20) | – | 6859 | 38107 | 1267 |
| step_bound(10) | – | 2846 | 8879 | 845 |

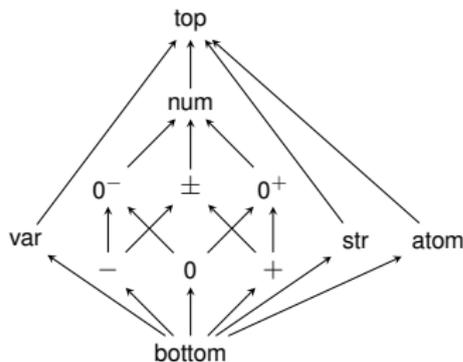
Performance Evaluation (Number of Answers)

Different answer management strategies



Performance evaluation – Tabling vs TCLP(Lat)

Simple **Abstract Interpreter** using sign abstract domain.



Two versions of Abstract Interpreter:

Tabling uses variant tabling.

TCLP(Lat) Uses entailment check to suspend.
Computation saved by reusing results from previous, more general, call.

| | Tabling | TCLP(Lat) |
|---------------------|---------|-----------|
| analyze(takeuchi/9) | 31.44 | 8.09 |
| analyze(takeuchi/7) | 13.75 | 5.85 |
| analyze(takeuchi/4) | 2.42 | 3.12 |

Constraint ASP Without Grounding

Motivation

ASP + constraints: *grounding phase* an issue since ranges of (constrained) variables may be *infinite*.

- Unbound range: $x \# > 0$ in \mathbb{IN}
- Bound range, but dense domain: $x \# > 0 \wedge x \# < 1$ in \mathbb{Q}

Current CASP systems (e.g., EZCSP [Balduccini and Lierler 2017] and clingo[DL/LP] [Janhunen et al. 2017]) *limit* (some of):

- Admissible constraint domains.
- Where constraints can appear.
- Type / number of models computed.

s(CASP): Main Points

- Adds constraints to s(ASP) [Marple et al. 2017], a **top-down** execution model that avoids the *grounding* phase.
- Is implemented with a goal-driven interpreter written in **Ciao Prolog**.
 - The execution of a program starts with a *query*.
 - Each answer provides the *mgu* of a successful derivation, its justification, and the *relevant* (partial) stable model.
- **Retains** variables and constraints during the execution and in the model.



<https://ciao-lang.org>

<https://gitlab.software.imdea.org/joaquin.arias/sCASP>

Background: s(ASP) [Marple et al. 2017]

- s(ASP) computes **constructive negation**: `not p(X)` returns in `X` the values for which `p(X)` fails.
 - Negated atoms are resolved against **dual rules** synthesized applying De Morgan's laws to Clark's completion of the original program.
- The construction of dual rules need two new operators:
 - **Disequality** (negation of the unification).
 - **Universal quantifier** (in the body of the clauses).
- To ensure that global constraints and consistency rules hold, **NMR-check rules** are synthesized and executed.
- The resulting program is executed by the s(ASP) interpreter which:
 - Carries around **explicitly** unification and disequality constraints.
 - Detects and handles different types of **loops**.

Background: Compilation of the Dual (Example)

- The *Dual* of a predicate P is another predicate that succeeds for the cases where P would have failed:
 - Clark completion.
 - Negation of \leftrightarrow .
 - Rearrangement of atoms.
 - Introduction of \neq and \forall .

Given the predicate:

```

1 p(0).
2 p(X) :- q(X), not t(X,Y).

```

Its dual rules are:

```

1 not p(X) :- not p1(X), not p2(X).
2 not p1(X) :- X \= 0.
3 not p2(X) :-
4     forall(Y, not p2_(X,Y)).
5 not p2_(X,Y) :- not q(X).
6 not p2_(X,Y) :- q(X), t(X,Y).

```

Background: Compilation of the Dual (Example)

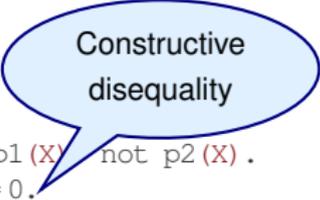
- The *Dual* of a predicate P is another predicate that succeeds for the cases where P would have failed:
 - Clark completion.
 - Negation of \leftrightarrow .
 - Rearrangement of atoms.
 - Introduction of \neq and \forall .

Given the predicate:

```
1 p(0).
2 p(X) :- q(X), not t(X,Y).
```

Its dual rules are:

```
1 not p(X) :- not p1(X) not p2(X).
2 not p1(X) :- X \= 0.
3 not p2(X) :-
4   forall(Y, not p2_(X,Y)).
5 not p2_(X,Y) :- not q(X).
6 not p2_(X,Y) :- q(X), t(X,Y).
```



Constructive disequality

Background: Compilation of the Dual (Example)

- The *Dual* of a predicate P is another predicate that succeeds for the cases where P would have failed:
 - Clark completion.
 - Negation of \leftrightarrow .
 - Rearrangement of atoms.
 - Introduction of \neq and \forall .

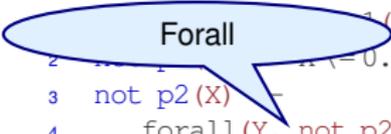
Given the predicate:

```

1 p(0).
2 p(X) :- q(X), not t(X,Y).

```

Its dual rules are:



```

1 not p1(X), not p2(X).
2 forall(X, not p1(X)).
3 not p2(X) :-
4   forall(Y, not p2_(X,Y)).
5 not p2_(X,Y) :- not q(X).
6 not p2_(X,Y) :- q(X), t(X,Y).

```

Background: Compilation of the Dual (Example)

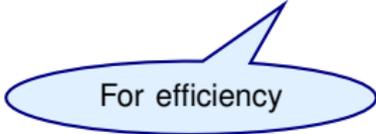
- The *Dual* of a predicate P is another predicate that succeeds for the cases where P would have failed:
 - Clark completion.
 - Negation of \leftrightarrow .
 - Rearrangement of atoms.
 - Introduction of \neq and \forall .

Given the predicate:

```
1 p(0).
2 p(X) :- q(X), not t(X,Y).
```

Its dual rules are:

```
1 not p(X) :- not p1(X), not p2(X).
2 not p1(X) :- X \= 0.
3 not p2(X) :-
4   forall(Y, not p2_(X,Y)).
5 not p2_(X,Y) :- not q(X).
6 not p2_(X,Y) :- q(X), t(X,Y).
```



For efficiency

Background: Compilation of the NMR-check (Example)

Given the consistency rule:

$$\forall \vec{x} (p(\vec{x}) \leftarrow \exists \vec{y} B \wedge \neg p(\vec{x}))$$

Any model should satisfy:

$$\forall \vec{x} \forall \vec{y} (\neg B \vee p(\vec{x}))$$

Background: Compilation of the NMR-check (Example)

Given the consistency rule:

$$\forall \vec{x} (p(\vec{x}) \leftarrow \exists \vec{y} B \wedge \neg p(\vec{x}))$$

Any model should satisfy:

$$\forall \vec{x} \forall \vec{y} (\neg B \vee p(\vec{x}))$$

1 `p(X) :- q(X,Y), ..., not p(X).`

1 `chk_1(X) :- forall(Y, not chk_1_(X,Y)).`

2 `not chk_1_(X,Y) :- not q(X,Y).`

3 `...`

4 `not chk_1_(X,Y) :- q(X,Y), ..., p(X).`

Background: Compilation of the NMR-check (Example)

Given the consistency rule:

$$\forall \vec{x} (p(\vec{x}) \leftarrow \exists \vec{y} B \wedge \neg p(\vec{x}))$$

```
1 p(X) :- q(X,Y), ..., not p(X).
```

$$\perp \leftarrow \exists \vec{x} \neg r(\vec{x})$$

```
1 :- not r(X).
```

Any model should satisfy:

$$\forall \vec{x} \forall \vec{y} (\neg B \vee p(\vec{x}))$$

```
1 chk_1(X) :- forall(Y, not chk_1_(X,Y)).
2 not chk_1_(X,Y) :- not q(X,Y).
3 ...
4 not chk_1_(X,Y) :- q(X,Y), ..., p(X).
```

$$\forall \vec{x} r(\vec{x})$$

```
1 chk_2 :- forall(X, r(X)).
```

Background: Compilation of the NMR-check (Example)

Given the consistency rule:

$$\forall \vec{x} (p(\vec{x}) \leftarrow \exists \vec{y} B \wedge \neg p(\vec{x}))$$

```
1 p(X) :- q(X,Y), ..., not p(X).
```

Any model should satisfy:

$$\forall \vec{x} \forall \vec{y} (\neg B \vee p(\vec{x}))$$

```
1 chk_1(X) :- forall(Y, not chk_1_(X,Y)).
2 not chk_1_(X,Y) :- not q(X,Y).
3 ...
4 not chk_1_(X,Y) :- q(X,Y), ..., p(X).
```

$$\perp \leftarrow \exists \vec{x} \neg r(\vec{x})$$

```
1 :- not r(X).
```

$$\forall \vec{x} r(\vec{x})$$

```
1 chk_2 :- forall(X, r(X)).
```

To ensure that each NMR-check rule is satisfied, the compiler adds the rule:

```
nmr_check :- forall(X,chk_1(X)), chk_2, ...
```

Background: \neq and `forall (X, Goal)`

Explanation delayed.

Background: Handling Loops (Goal eventually Calling Itself)

s(ASP) interpreter checks existence of different types of loops:

Background: Handling Loops (Goal eventually Calling Itself)

s(ASP) interpreter checks existence of different types of loops:

- **Odd loop over negation**: recursion with an odd number of intervening negations: fails (avoid inconsistencies).

```

1  p(X) :- q(X), not p(X).      2  q(a).
                                ?- p(a).
                                no
    
```

Background: Handling Loops (Goal eventually Calling Itself)

s(ASP) interpreter checks existence of different types of loops:

- **Odd loop over negation:** recursion with an odd number of intervening negations: fails (avoid inconsistencies).

```

1  p(X) :- q(X), not p(X).      2  q(a).
                                ?- p(a).
                                no
  
```

- **Even loop over negation:** Id. with even (non zero) number of intervening negations. It generates multiple models.

```

1  p(X) :- not q(X).           2  q(X) :- not p(X).
                                ?- p(a).
                                {p(a), not q(a)}
  
```

Background: Handling Loops (Goal eventually Calling Itself)

s(ASP) interpreter checks existence of different types of loops:

- **Odd loop over negation:** recursion with an odd number of intervening negations: fails (avoid inconsistencies).

```
1 p(X) :- q(X), not p(X).      2 q(a).
                                ?- p(a).
                                no
```

- **Even loop over negation:** Id. with even (non zero) number of intervening negations. It generates multiple models.

```
1 p(X) :- not q(X).           2 q(X) :- not p(X).
                                ?- p(a).
                                {p(a), not q(a)}
```

- **Positive loops:** No intervening negations. Fail if calls match (as in tabling).

```
1 p(X) :- ..., p(X).
                                ?- p(a).
                                no
```

s(CASP): Design and Implementation

Main contributions w.r.t. s(ASP) are:

- Re-implemented [interpreter](#): Prologs takes care of environment (e.g., the behavior of variables).

s(CASP): Design and Implementation

Main contributions w.r.t. s(ASP) are:

- Re-implemented [interpreter](#): Prologs takes care of environment (e.g., the behavior of variables).
- (Simple) constraint solver for disequality, [CLP\(\$\neq\$ \)](#), using attributed variables.

s(CASP): Design and Implementation

Main contributions w.r.t. s(ASP) are:

- Re-implemented [interpreter](#): Prologs takes care of environment (e.g., the behavior of variables).
- (Simple) constraint solver for disequality, [CLP\(\$\neq\$ \)](#), using attributed variables.
- Generic CLP interface and an extended compiler to [plug in](#) constraint solvers.

s(CASP): Design and Implementation

Main contributions w.r.t. s(ASP) are:

- Re-implemented [interpreter](#): Prologs takes care of environment (e.g., the behavior of variables).
- (Simple) constraint solver for disequality, [CLP\(\$\neq\$ \)](#), using attributed variables.
- Generic CLP interface and an extended compiler to [plug in](#) constraint solvers.
- Design and implementation of [C-forall](#) — generalizes original [forall](#) algorithm, to support constraints in arbitrary domains.

s(CASP): Design and Implementation

Main contributions w.r.t. s(ASP) are:

- Re-implemented [interpreter](#): Prologs takes care of environment (e.g., the behavior of variables).
- (Simple) constraint solver for disequality, [CLP\(\$\neq\$ \)](#), using attributed variables.
- Generic CLP interface and an extended compiler to [plug in](#) constraint solvers.
- Design and implementation of [C-forall](#) — generalizes original [forall](#) algorithm, to support constraints in arbitrary domains.

| | s(CASP) | s(ASP) |
|---------------|----------------|---------------|
| hanoi(8,T) | 1,528 | 13,297 |
| queens(4,Q) | 1,930 | 20,141 |
| One hamicycle | 493 | 3,499 |
| Two hamicycle | 3,605 | 18,026 |

Run time (ms) comparison s(CASP) vs. s(ASP).

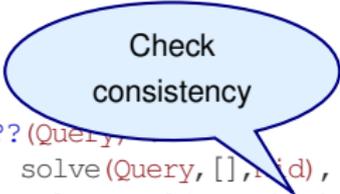
s(CASP): The Interpreter

```

1  ??(Query) :-
2    solve(Query, [], Mid),
3    solve_goal(nmr_check, Mid, Out),
4    output_just_model(Out).
5
6  solve([], In, ['$success' | In]).
7  solve([Goal|Gs], In, Out) :-
8    solve_goal(Goal, In, Mid),
9    solve(Gs, Mid, Out).
10 solve_goal(Goal, In, Out) :-
11   user_defined(Goal), !,
12   check_loops(Goal, In, Out).
13 solve_goal(Goal, In, Out) :-
14   Goal = forall(V, FGoal), !,
15   c_forall(V, FGoal, In, Out).
16 solve_goal(Goal, In, Out) :-
17   call(Goal),
18   Out = ['$success', Goal | In].
  
```

Figure: (Very abridged) Code of the s(CASP) interpreter.

s(CASP): The Interpreter



Check consistency

```

1  ??(Query,
2     solve(Query, [], Mid),
3     solve_goal(nmr_check, Mid, Out),
4     output_just_model(Out)).
5
6  solve([], In, ['$success' | In]).
7  solve([Goal|Gs], In, Out) :-
8     solve_goal(Goal, In, Mid),
9     solve(Gs, Mid, Out).
10 solve_goal(Goal, In, Out) :-
11     user_defined(Goal), !,
12     check_loops(Goal, In, Out).
13 solve_goal(Goal, In, Out) :-
14     Goal = forall(V, FGoal), !,
15     c_forall(V, FGoal, In, Out).
16 solve_goal(Goal, In, Out) :-
17     call(Goal),
18     Out = ['$success', Goal | In].

```

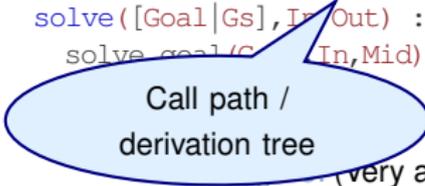
Figure: (Very abridged) Code of the s(CASP) interpreter.

s(CASP): The Interpreter

```

1  ??(Query) :-
2     solve(Query, [], Mid),
3     solve_goal(nmr_check, Mid, Out),
4     output_just_model(Out).
5
6  solve([], In, ['$success' | In]).
7  solve([Goal|Gs], In, Out) :-
8     solve_goal(Goal, In, Mid),
9
10     solve_goal(Goal, In, Out) :-
11         user_defined(Goal), !,
12         check_loops(Goal, In, Out).
13     solve_goal(Goal, In, Out) :-
14         Goal = forall(V, FGoal), !,
15         c_forall(V, FGoal, In, Out).
16     solve_goal(Goal, In, Out) :-
17         call(Goal),
18         Out = ['$success', Goal | In].

```



Call path /
derivation tree

(very abridged) Code of the s(CASP) interpreter.

s(CASP): The Interpreter

```

1  ??(Query) :-
2    solve(Query, [], Mid),
3    solve_goal(nmr_check, Mid, Out),
4    output_just_model(Out).
5
6  solve([], In, ['$success' | In]).
7  solve([Goal|Gs], In, Out) :-
8    solve_goal(Goal, In, Mid),
9    solve(Gs, Mid, Out).
10 solve_goal(Goal, In, Out) :-
11   user_define(Goal), !,
12   check_loops(Goal, In, Out).
13 solve_goal(Goal, In, Out) :-
14   Goal = forall(V, FGoal), !,
15   c_forall(V, FGoal, In, Out).
16 solve_goal(Goal, In, Out) :-
17   call(Goal),
18   Out = ['$success', Goal | In].

```

Detect and handle loops

Figure: (Very abridged) Code of the s(CASP) interpreter.

```

...
check_loops(Goal, In, Out) :-
  pr_rule(Goal, Body),
  solve(Body, [Goal | In], Out).

```

s(CASP): The Interpreter

```

1  ??(Query) :-
2     solve(Query, [], Mid),
3     solve_goal(nmr_check, Mid, Out),
4     output_just_model(Out).
5
6  solve([], In, ['$success' | In]).
7  solve([Goal|Gs], In, Out) :-
8     solve_goal(Goal, In, Mid),
9     solve(Gs, Mid, Out).
10 solve_goal(Goal, In, Out) :-
11     user_defined(Goal), !,
12     k_loops(Goal, In, Out).
13 solve_goal(Goal, In, Out) :-
14     Goal = forall(V, FGoal), !,
15     c_forall(V, FGoal, In, Out).
16 solve_goal(Goal, In, Out) :-
17     call(Goal),
18     Out = ['$success', Goal | In].

```

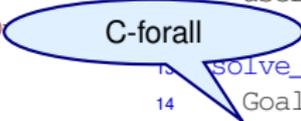


Figure: (Very abridged) Code of the s(CASP) interpreter.

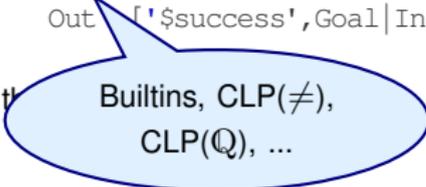
s(CASP): The Interpreter

```

1  ??(Query) :-
2    solve(Query, [], Mid),
3    solve_goal(nmr_check, Mid, Out),
4    output_just_model(Out).
5
6  solve([], In, ['$success' | In]).
7  solve([Goal|Gs], In, Out) :-
8    solve_goal(Goal, In, Mid),
9    solve(Gs, Mid, Out).
10 solve_goal(Goal, In, Out) :-
11   user_defined(Goal), !,
12   check_loops(Goal, In, Out).
13 solve_goal(Goal, In, Out) :-
14   Goal = forall(V, FGoal), !,
15   c_forall(V, FGoal, In, Out).
16 solve_goal(Goal, In, Out) :-
17   call(Goal),
18   Out = ['$success', Goal | In].

```

Figure: (Very abridged) Code of the



Builtins, CLP(\neq),
CLP(Q), ...

s(CASP): `c_forall (X, Goal)`

Check if `Goal` is true for all values in the constraint domain of `X`.

Intuition: Narrow the constraint store C_i under which `Goal` is executed by selecting an answer A_i and removing from C_i the values of `X` covered by A_i .

A_x is the projection of A onto `X`.

$A_{\bar{x}}$ Id. onto the set of variables in `Goal` that are not `X`.

s(CASP): `c_forall (X, Goal)`

Check if `Goal` is true for all values in the constraint domain of `X`.

Intuition: Narrow the constraint store C_i under which `Goal` is executed by selecting an answer A_i and removing from C_i the values of X covered by A_i .

$A_{\bar{X}}$ is the projection of A onto \bar{X} .

$A_{\bar{X}}$ Id. onto the set of variables in `Goal` that are not X .

Algorithm (simplified):

- If `Goal` succeeds with answer A_i under C_i , there are two possibilities:
 - $A_{i,X} \equiv C_{i,X}$ then succeed.
 - $A_{i,X} \sqsubset C_{i,X}$ then re-execute `Goal` under $C_{i+1} = C_i \wedge A_{i,\bar{X}} \wedge \neg A_{i,X}$.
- If `Goal` fails, then fail.

Note 1: `c_forall/2` takes care of **disjunctions** generated by $\neg A_{i,X}$
 (Constraints solvers usually cannot handle them natively.)

$s(\text{CASP}): c_forall(X, \text{Goal})$

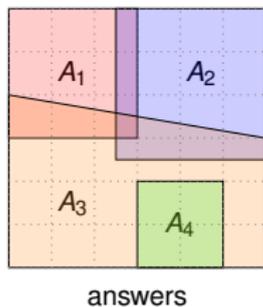
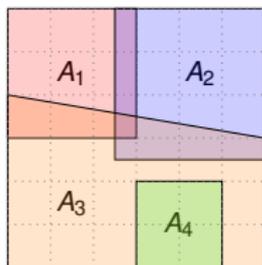
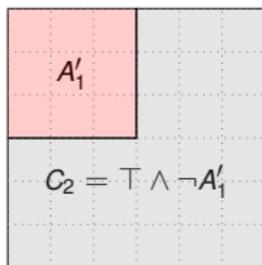


Figure: A *C-forall* evaluation that succeeds.

s(CASP): $c_forall(X, Goal)$



answers



(a)

Figure: A *C-forall* evaluation that succeeds.

s(CASP): $c_forall(X, Goal)$

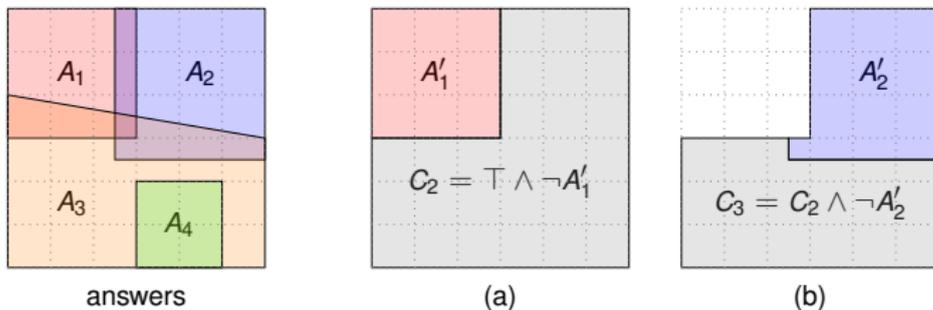


Figure: A C -forall evaluation that succeeds.

s(CASP): $c_forall(X, Goal)$

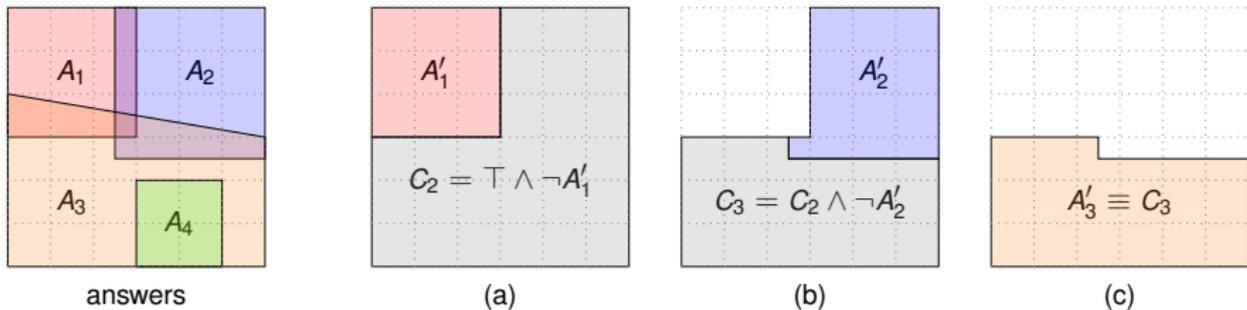


Figure: A *C-forall* evaluation that succeeds.

s(CASP): `c_forall (X, Goal)`

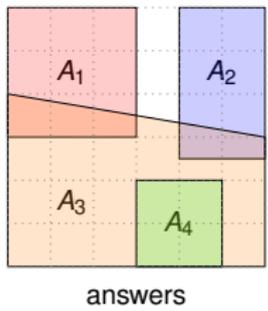


Figure: A *C-forall* evaluation that fails.

s(CASP): `c_forall (X, Goal)`

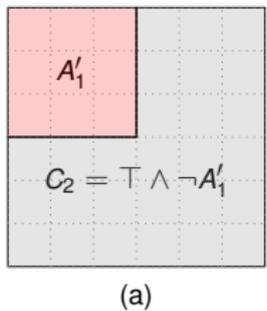
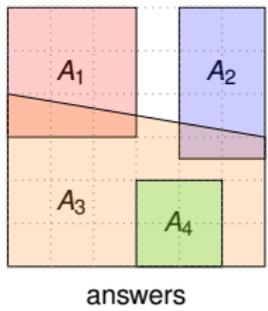
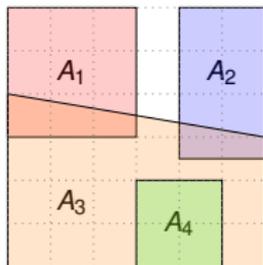
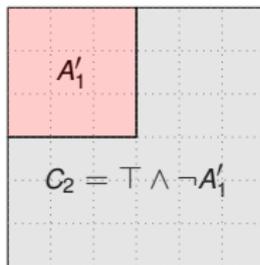


Figure: A *C-forall* evaluation that fails.

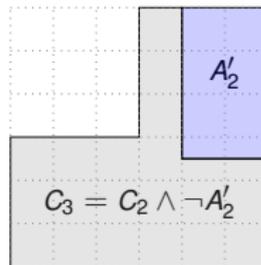
s(CASP): `c_forall (X, Goal)`



answers



(a)



(b)

Figure: A *C-forall* evaluation that fails.

s(CASP): $c_forall(X, Goal)$

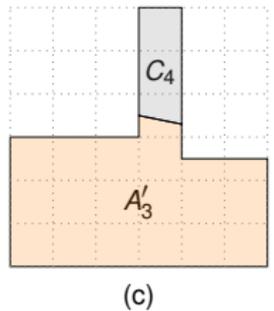
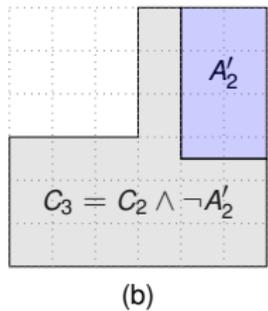
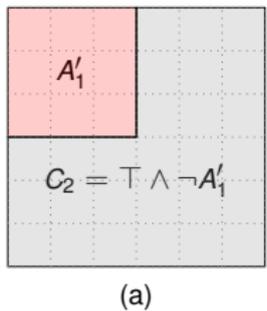
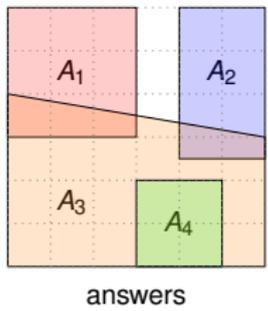
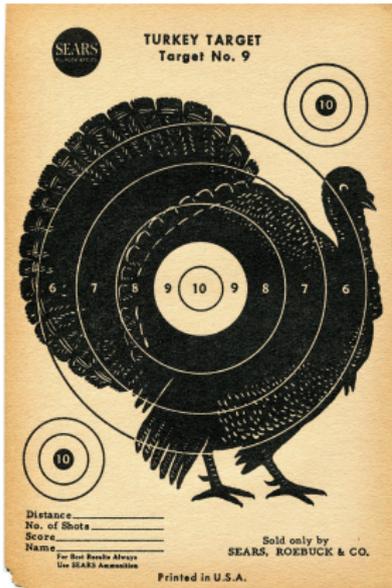


Figure: A *Cforall* evaluation that fails.

s(CASP): Yale Shooting Scenario [Janhunen et al. 2017]



- There is a turkey, a gun, and three possible actions: wait, load, shoot.
- Initially: turkey alive, gun unloaded.
- The turkey will die if we load and shoot within 35 minutes. Otherwise, the gun powder is spoiled.
- We are not allowed to shoot in the first 35 minutes.
- We want a plan to kill the turkey within 100 minutes.

s(CASP): Yale Shooting Scenario [Janhunen et al. 2017]

```

1 duration(load,25).
2 duration(shoot,5).
3 duration(wait,36).
4 spoiled(T_Armed):- T_Armed #> 35.
5 prohibited(shoot,Time):- Time #< 35.
6
7 holds(0, State, []):- init(State).
8 holds(F_Time, F_State, [Action|As]):-
9     F_Time #> 0,
10    F_Time #= P_Time + Duration,
11    duration(Action, Duration),
12    not prohibited(Action, F_Time),
13    trans(Action, P_State, F_State),
14    holds(P_Time, P_State, As).
15
16
17 init(st(alive,unloaded,0)).
18
19 trans(load, st(alive,_,_),
20        st(alive,loaded,0)).
21
22 trans(wait, st(alive,Gun,P_Armed),
23        st(alive,Gun,F_Armed)):-
24    F_Armed #= P_Armed + Duration,
25    duration(wait,Duration).
26
27 trans(shoot, st(alive,loaded,T_Armed),
28        st(dead,unloaded,0)):-
29    not spoiled(T_Armed).
30
31 trans(shoot, st(alive,loaded,T_Armed),
32        st(alive,unloaded,0)):-
33    spoiled(T_Armed).

```

s(CASP) code for the Yale Shooting problem

s(CASP): Yale Shooting Scenario [Janhunen et al. 2017]

Restrictions
as constraints

```

1 duration(load,25).
2 duration(shoot,5).
3 duration(wait,36).
4 spoiled(T_Armed):- T_Armed #> 35.
5 prohibited(shoot,Time):- Time #< 35.
6
7 holds(0, State, []):- init(State).
8 holds(F_Time, F_State, [Action|As]):-
9     F_Time #> 0,
10    F_Time #= P_Time + Duration,
11    duration(Action, Duration),
12    not prohibited(Action, F_Time),
13    trans(Action, P_State, F_State),
14    holds(P_Time, P_State, As).
17 trans(load, st(alive,_,_),
18         st(alive,loaded,0)).
19 trans(wait, st(alive,Gun,P_Armed),
20         st(alive,Gun,F_Armed)):-
21     F_Armed #= P_Armed + Duration,
22     duration(wait,Duration).
23 trans(shoot, st(alive,loaded,T_Armed),
24         st(dead,unloaded,0)):-
25     not spoiled(T_Armed).
26 trans(shoot, st(alive,loaded,T_Armed),
27         st(alive,unloaded,0)):-
28     spoiled(T_Armed).

```

s(CASP) code for the Yale Shooting problem

s(CASP): Yale Shooting Scenario [Janhunen et al. 2017]

```

1 duration(load,25).
2 duration(shoot,5).
3 duration(wait,36).
4 spoiled(T_Armed):- T_Armed #> 35.
5 prohibited(shoot,Time):- Time #< 35.
6
7 holds(0, State, []):- init(State).
8 holds(0, Time, F_State, [Action|As]):-
9     0,
10    F_Time #= P_Time + Duration,
11    duration(Action, Duration),
12    not prohibited(Action, F_Time),
13    trans(Action, P_State, F_State),
14    holds(P_Time, P_State, As).
15
16 trans(load, st(alive,_,_),
17        st(alive,loaded,0)).
18
19 trans(wait, st(alive,Gun,P_Armed),
20          st(alive,Gun,F_Armed)):-
21    F_Armed #= P_Armed + Duration,
22    duration(wait, Duration).
23
24 trans(shoot, st(alive,loaded,T_Armed),
25         st(dead,unloaded,0)):-
26    not spoiled(T_Armed).
27
28 trans(shoot, st(alive,loaded,T_Armed),
29        st(alive,unloaded,0)):-
30    spoiled(T_Armed).

```

Restrictions
as constraints

Negation

Negation

s(CASP) code for the Yale Shooting problem

s(CASP): Yale Shooting Scenario [Janhunen et al. 2017]

```

1  duration(load,25) .
2  duration(shoot,5) .
3  duration(wait,36) .
4  spoiled(T_Armed):- T_Armed #> 35.
5  prohibited(shoot,Time):- Time #< 35.
6
7  holds(0, State, []):- init(State) .
8  holds(F_Time, F_State, [Action|As]):-
9      F_Time #> 0,
10     F_Time #= P_Time + Duration,
11     duration(Action, Duration) ,
12     not prohibited(Action, F_Time) ,
13     trans(Action, P_State, F_State) ,
14     holds(P_Time, P_State, As) .
15  init(st(alive,unloaded,0)) .
16
17  trans(load, st(alive,_,_) ,
18         st(alive,loaded,0)) .
19  trans(wait, st(alive,Gun,P_Armed) ,
20         st(alive,Gun,F_Armed)):-
21     F_Armed #= P_Armed + Duration,
22     duration(wait,Duration) .
23  trans(shoot, st(alive,loaded,T_Armed) ,
24         st(dead,unloaded,0)):-
25     not spoiled(T_Armed) .
26  trans(shoot, st(alive,loaded,T_Armed) ,
27         st(alive,unloaded,0)):-
28     spoiled(T_Armed) .

```

s(CASP) code for the Yale Shooting problem

```
% holds(Time, st(Turkey,Gun,Time_Armed) , Plan)
```

s(CASP): Yale Shooting Scenario [Janhunen et al. 2017]

```

1  duration(load,25).
2  duration(shoot,5).
3  duration(wait,36).
4  spoiled(T_Armed):- T_Armed #> 35.
5  prohibited(shoot,Time):- Time #< 35.
6
7  holds(0, State, []):- init(State).
8  holds(F_Time, F_State, [Action|As]):-
9      F_Time #> 0,
10     F_Time #= P_Time + Duration,
11     duration(Action, Duration),
12     not prohibited(Action, F_Time),
13     trans(Action, P_State, F_State),
14     holds(P_Time, P_State, As).
15  init(st(alive,unloaded,0)).
16
17  trans(load, st(alive,_,_),
18         st(alive,loaded,0)).
19  trans(wait, st(alive,Gun,P_Armed),
20         st(alive,Gun,F_Armed)):-
21     F_Armed #= P_Armed + Duration,
22     duration(wait,Duration).
23  trans(shoot, st(alive,loaded,T_Armed),
24         st(dead,unloaded,0)):-
25     not spoiled(T_Armed).
26  trans(shoot, st(alive,loaded,T_Armed),
27         st(alive,unloaded,0)):-
28     spoiled(T_Armed).

```

s(CASP) code for the Yale Shooting problem

```

?- Time #< 100, holds(Time, st(dead,_,_), Plan).

```

s(CASP): Yale Shooting Scenario [Janhunen et al. 2017]

```
?- ?? [Time #< 100, holds (Time, st (dead, _, _), Plan)] .
```

```
Time=55, Plan=[shoot, load, load]
```

```
Time=66, Plan=[shoot, load, wait]
```

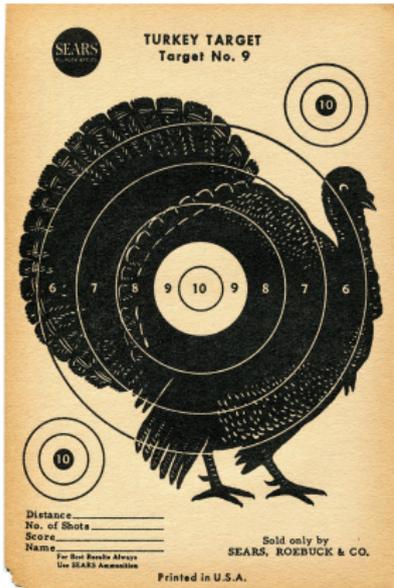
```
Time=80, Plan=[shoot, load, load, load]
```

```
Time=91, Plan=[shoot, load, load, wait]
```

```
Time=91, Plan=[shoot, load, wait, load]
```

```
Time=96, Plan=[shoot, load, shoot, wait, load]
```

s(CASP): Yale Shooting Scenario extended



Extensions:

- Time is dense \rightarrow intervals have infinite # of elements.
- There is a second gun and initially only one of them is loaded.
- We cannot shoot in the first 35 minutes only if our gun is initially unloaded.

s(CASP): Yale Shooting Scenario extended

```

1 duration(load,25).
2 duration(shoot,D):- D #> 5, D #< 15/2.
3 duration(wait,36).
4 spoiled(T_Armed):- T_Armed #> 35.
5 prohibited(shoot,Time):-
6     Time #< 35, gun(unloaded).
7
8 holds(0, State, [1]).
9 holds(F_Time, State):-
10     F_Time #> 0,
11     F_Time #= P_Time + Duration,
12     duration(Action, Duration),
13     not prohibited(Action, F_Time),
14     trans(Action, F_Time),
15     holds(P_Time, P_State, As).
16
17 init(st(alive,Gun,0)) :- gun(Gun).
18
19 trans(load, st(alive,_,_),
20         st(alive,loaded,0)).
21 trans(shoot, st(alive,loaded,T_Armed),
22         st(dead,unloaded,0)):-
23     not spoiled(T_Armed).
24
25 trans(wait, st(alive,loaded,T_Armed),
26         st(alive,unloaded,0)):-
27     not s_gun(loaded).
28
29 gun(loaded) :- not s_gun(loaded).
30 gun(unloaded) :- not gun(loaded).
31 s_gun(unloaded) :- not s_gun(loaded).

```

Restriction

Interval in a dense domain

Two possible worlds

Initial state

s(CASP) code for the extended and updated Yale Shooting problem.

s(CASP): Other Examples

Stream Data Reasoning: constraints and goal-directed strategy make it possible to answer queries without evaluating the complete stream database.

```

1  valid_stream(Pr,Data) :-
2      stream(Pr,Data),
3      not cancelled(Pr,Data) .
4
5  cancelled(PrLo,DataLo) :-
6      PrHi #> PrLo,
7      stream(PrHi,DataHi),
8      incompot(DataLo,DataHi) .
  
```

s(CASP): Other Examples

Stream Data Reasoning: constraints and goal-directed strategy make it possible to answer queries without evaluating the complete stream database.

```

1  valid_stream(Pr,Data) :-
2      stream(Pr,Data),
3      not cancelled(Pr,Data).
4
5  cancelled(PrLo,DataLo) :-
6      PrHi #> PrLo,
7      stream(PrHi,DataHi),
8      incompot(DataLo,DataHi).

```

Traveling Salesman Problem: s(CASP) encoding is more compact than CLP and constraints (over dense domains) can appear as part of the model.

```

?- travel_path(b,Length,Cycle).

{ cycle_dist(b,c,31/10), cycle_dist(c,d,A) {A #> 8, A #< 21/2},
  cycle_dist(d,a,1), cycle_dist(a,b,1) }

```

General Thoughts

- Constraint + ASP subject of a lot of research.
- Constraints work better with the notion of variables!
 - E.g., intensional description of sets.
- Traditional ASP evaluation shares some points with bottom-up.
 - I.e., both do not in principle use variables.
- But smart top-down evaluation (tabling) achieves results similar to bottom up.
 - And variables and relationships can be used.
- Can the case for ASP be similar?

Bibliography I

- Balduccini, M. and Lierler, Y. (2017). Constraint Answer Set Solver EZCSP and why Integration Schemas Matter. *Theory and Practice of Logic Programming*, 17(4):462–515.
- Holzbaur, C. (1995). OFAI CLP(Q,R) Manual, Edition 1.3.3. Technical Report TR-95-09, Austrian Research Institute for Artificial Intelligence, Vienna.
- Janhunen, T., Kaminski, R., Ostrowski, M., Schellhorn, S., Wanko, P., and Schaub, T. (2017). Clingo goes Linear Constraints over Reals and Integers. *TPLP*, 17(5-6):872–888.
- Marple, K., Salazar, E., and Gupta, G. (2017). Computing Stable Models of Normal Logic Programs Without Grounding. *CoRR*, abs/1709.00501.
- Toman, D. (1997). Memoing Evaluation for Constraint Extensions of Datalog. *Constraints*, 2(3/4):337–359.