ID2204: Constraint Programming

Constraint Propagation

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Christian Schulte

cschulte@kth.se

Software and Computer Systems
School of Information and Communication Technology
KTH – Royal Institute of Technology
Stockholm, Sweden



KTH Information and Communication Technology

Naïve Constraint Propagation

Naïve Constraint Propagation

Looking for

```
propagate : M \times S \rightarrow S
```

M = set of all models S = set of all stores

performing constraint propagation

- start from some initial store
- return store on which all propagation has been performed
- ignore efficiency, focus on principle idea

Naïve Propagation Function

V = set of decision variables U = universe = common domain $\text{propagate}((V, U, P), s) \stackrel{P}{=} \text{ set of propagators of model}$ s = store to start from s := p(s); return s;

- What is returned as result?
- Does it terminate?

Termination

Consider store s_i at i-th iteration of loop with s_o initial store

$$S_{j+1} < S_j$$

- That is, s_i form strictly decreasing sequence: cannot be infinite
 - remember: (S,<) is well-founded!</p>
- Loop terminates!

Result Computed

Assume propagate ((V, U, P), s) = s'

$$sol((V,U,P),s) = sol((V,U,P),s')$$

no solutions removed

for all
$$p \in P$$
: $p(s') = s'$

no further propagation possible

the weakest simultaneous fixpoint: unique (by mono.)

NB: a strongest simultaneous fixpoint would be a solution (hence not unique), which would violate solution preservation

Weakest Simultaneous Fixpoint

Assume propagate ((V, U, P), s) = s'Then

s' weakest sim. fixpoint with s'≤ s that is

for all
$$p \in P$$
 $p(s') = s'$

- clear, follows from termination of loop weakest fixpoint?
- any other fixpoint is stronger

Why Naïve?

- Always searches all propagators of model for propagator which can contract strictly
 - maintain propagators which are known to have fixpoint computed
 - might have to find out by having propagators which do no contraction
 - take variables into account which connect two propagators

Realistic Constraint Propagation

Improving Propagation

- Idea: propagator narrows domain of some (few) variables
 - re-propagate only propagators sharing modified variables
- Maintain a set of "dirty" propagators
 - not known whether at fixpoint for current store
 - all other propagators have fixpoint computed
 - only propagate "dirty" propagators

Propagator Variables

- Variables var(p) of propagator p
 - variables of interest
- No input considered from other variables
- No output computed on other variables

Variable Dependencies

- No output on other variables for all s∈S, for all x∈(V-var(p)) p(s)(x)=s(x)
- No input from other variables for all s₁, s₂ ∈ S if (for all x∈var(p): s₁(x)=s₂(x)), then (for all x∈var(p): p(s₁)(x)=p(s₂)(x))

Idea: Improved Propagation

- maintain set N of "dirty" propagators
- choose propagator to run next from N and remove from N
- compute modified variables
- add propagators sharing variables with modified variables to N
 - including the running propagator!
 (as propagators need not be idempotent)

Improved Propagation

```
propagate((V, U, P), s_0)
                s := s_0; N := P;
                while N \neq \emptyset do
                        choose p \in N;
                        s' := p(s); N := N - \{p\};
(dependent propagators) MV := \{ x \in V \mid s(x) \neq s'(x) \}; \text{ (modified variables)} 
DP := \{ q \in P \mid \text{ exists } x \in \text{var}(q) : x \in MV \}; 
N := N \cup DP; \text{ (maintain set N of dirty propagators)}
                        s := s':
                return s;
```

Questions

- What does it compute
 - does it compute simultaneous fixpoint?
 - the weakest?
 - important: loop invariant

- Termination?
 - stores are not any longer strictly stronger

Summary:

Propagators

- compute with stores
- are contracting and monotonic
- maintain solutions
- strong enough to decide for assignments

Naïve propagation

- terminates
- computes weakest simultaneous fixpoint
- Dependency directed propagation
 - if variables shrink, rerun dependent propagators only
 - Computes weakest simultaneous fixpoint
 - terminates

Improving Propagation Further

General Idea

- Essential: knowledge on fixpoint for a propagator
- So far: only implicit knowledge
- Here: let us make knowledge explicit
 - propagators provide information

We Are Done! What Now?

Suppose the following propagator

$$p(s) = \{x \rightarrow (s(x) \cap \{1,2,3\})\}$$

- implements domain constraint x∈{1,2,3}
- After executing p once, no further execution needed:

if
$$s' \le p(s)$$
 then $p(s')=s'$

- We can safely delete p from model
 - otherwise, pointless re-execution!

Subsumed Propagators

Definition:

Propagator p subsumed by store s, iff

for all
$$s' \le s : p(s') = s'$$

- all stronger stores are fixpoints
- p entailed by s
- s subsumes p (s entails p)

Reminder: Propagator for ≤

■ Propagator p_{\leq} for $x \leq y$

$$p_{\leq}(s) = \{ x \rightarrow \{ n \in s(x) \mid n \leq \max(s(y)) \}, \\ y \rightarrow \{ n \in s(y) \mid n \geq \min(s(x)) \} \}$$

Example:

```
p_{<}= is subsumed by \{x:\{1,2,3\}, y:\{3,4,5\}\}. p_{<}= is not subsumed by \{x:\{1,2,3,4\}, y:\{3,4,5\}\}.
```

We Are Done! What Next?

After executing p_≤ on store s we have

$$p_{\leq}(p_{\leq}(s))=p_{\leq}(s)$$

- max(s(y)) does not change!
- min(s(x)) does not change!
- What happens: as var(p_≤)={x,y}, p_≤ is added to DP and hence to N
 - but: s' is fixpoint for p_{\leq}
 - no need to include in DP

First Attempt: Idempotent Functions

Definition:

■ A function $f \in X \rightarrow X$ is *idempotent*, if for all $x \in X$: f(f(x)) = f(x)

Very strong property for a propagator: required for all stores!

An example of a non-idempotent propagator is given in Example 2.9 on page 19 of Course Notes.

A domain consistent propagator is necessarily idempotent.

A bounds(Z) consistent propagator is not necessarily idempotent.

Second Attempt: Weak Idempotence

A function f ∈ X → X is idempotent on x ∈ X if x is a fixpoint of f:

$$f(f(x)) = f(x)$$

- statement on just one element
- For a propagator: if p is idempotent on s, it does not mean that p is idempotent on s' with s' ≤ s

How to Find Out?

Given store s and propagator p

- Does s subsume p?
 - try all s' < s: way toocostly</p>
- Is p idempotent on s?
 - apply p to s: that is what we tried
 to avoid in 1st place

Status Messages

Solution: propagator returns status and tells result

```
propagator p is function p \in S \rightarrow SM \times S with SM := \{fix, nofix, subsumed\}
```

Propagator with Status

Assume propagator *p* and store *s* if p(s) = (fix, s'), then s' is fixpoint for p if p(s) = (subsumed, s'), then s' subsumes p if p(s) = (nofix, s'), then no further knowledge always safe (as before)

Propagator for ≤ with Subsumption

Propagator $p_{<}$ for $x \le y$ $p_{<}(s) =$ if $max(s(x)) \le min(s(y))$ then (subsumed, s) else (fix, $\{x \rightarrow \{n \in s(x) \mid n \leq \max(s(y))\},\$ $v \rightarrow \{ n \in s(v) \mid n \geq \min(s(x)) \} \}$

but subsumption could also be tested for in the else case!

Propagator for ≤ with Subsumption (better version)

■ Propagator p_{\leq} for $x \leq y$ $ep_{\leq}(s) = \text{let } s' = p_{\leq}(s) \text{ in}$ $\text{if } \max(s'(x)) \leq \min(s'(y)) \text{ then}$ (subsumed, s') else (fixpt, s')

What to Return?

- Propagation function now also needs to return the set of propagators
 - in case of subsumption, propagators are removed

Improved Propagation

```
propagate((V, U, P), s_0)
    s := s_0; N := P;
    while N \neq \emptyset do
        choose p \in N;
         (sm,s'):=p(s); N:=N-\{p\};
         if sm=subsumed then P := P - \{p\}; end
         MV := \{ x \in V \mid s(x) \neq s'(x) \};
        DP := \{ q \in P \mid \text{exists } x \in \text{var}(q) : x \in MV \};
         if sm=fix then DP := DP - \{p\}; end
         N := N \cup DP;
         s := s';
    return (P,s);
```

Correctness

- Are the optimizations correct?
- How to prove:
 - invariant is still invariant
 - solutions remain the same
 - still computes the same

argument: fixpoints!

Propagation Events

Propagation Events

- Many propagators
 - simple to decide whether still at fixpoint for changed domain
 - based on how domain has changed
- How domain changes described by *propagation event* or just event

Propagator for \leq

■ Propagator p_{\leq} for $x \leq y$

$$p_{\leq}(s) = \{ x \rightarrow \{ n \in s(x) \mid n \leq \max(s(y)) \}, \\ y \rightarrow \{ n \in s(y) \mid n \geq \min(s(x)) \} \}$$

must be propagated only if max(s(y)) or min(s(x)) changes

Propagator for

■ Propagator p_{\neq} for $x \neq y$

$$p_{\neq}(s) =$$

$$\{x \rightarrow s(x) - \text{single}(s(y)), \\ y \rightarrow s(y) - \text{single}(s(x))\}$$
• where: $\text{single}(n) = n$
 $\text{single}(N) = \emptyset \text{ (otherwise)}$

must be propagated only if x or y become assigned

Events

Typical events

```
fix(x) x becomes assigned
```

- = min(x) minimum of x changes
- = max(x) maximum of x changes
- any(x) domain of x changes

Clearly overlap

- fix(x) occurs: min(x) or max(x) occur any(x) occurs
- min(x) or max(x) occur: any(x) occurs

Do not mix up the "fix" status message with the "fix(x)" propagation event.

Events on Store Change

```
events(s,s') =
\{ any(x) \mid s'(x) \subset s(x) \} \cup \{ min(x) \mid min s'(x) > min s(x) \} \cup \{ max(x) \mid max s'(x) < max s(x) \} \cup \{ fix(x) \mid |s'(x)|=1 and |s(x)|>1 \}
```

where *s*' ≤ *s*

Events: Example

Given stores

■
$$s = \{ x_1 \rightarrow \{1,2,3\}, x_2 \rightarrow \{3,4,5,6\}, x_3 \rightarrow \{0,1\}, x_4 \rightarrow \{7,8,10\} \}$$

■ $s' = \{ x_1 \rightarrow \{1,2\}, x_2 \rightarrow \{3,5,6\}, x_3 \rightarrow \{1\}, x_4 \rightarrow \{7,8,10\} \}$

■ Then events(s,s') =

```
{ max(x_1), any(x_1),
any(x_2),
fix(x_3), min(x_3), any(x_3)}
```

Events are Monotonic

• If $s'' \le s'$ and $s' \le s$ then events(s,s'') =events $(s,s') \cup \text{events}(s',s'')$

- Event occurs on change from s to s"
 - occurs on change from s to s', or
 - occurs on change from s' to s"

Event Sets: First Requirement

- Event set for propagator p: es(p)
 - for all stores s' and s with s' ≤ s and s'(x)= s(x) for all x∈ V-var(p)

```
if p(s)=s and p(s')\neq s' then es(p)\cap events(s,s')\neq \emptyset
```

if store s is fixpoint and changes to non-fixpoint s', then events from s to s' must be included in es(p)

Event Sets: Second Requirement

- Event set for propagator p: es(p)
 - for all stores s with p(p(s))≠p(s):
 es(p) ← events(s,p(s)) ≠∅
 - if propagator does not compute fixpoint on store s, then events from s to p(s) must be included in es(p)
 - does not occur for idempotent propagators (which compute fixpoints in one go)

Propagator for ≤

■ Propagator p_{\leq} for $x \leq y$

$$p_{\leq}(s) = \{ x \rightarrow \{ n \in s(x) \mid n \leq \max(s(y)) \}, \\ y \rightarrow \{ n \in s(y) \mid n \geq \min(s(x)) \} \}$$

- good one: $es(p_{\leq}) = \{ max(y), min(x) \}$
- but also: es(p≤) = { any(y), any(x)}

Propagator for \(\neq \)

Propagator p_≠ for x ≠ y

```
p_{\neq}(s) =
 \{ x \rightarrow s(x) - \text{single}(s(y)), 
 y \rightarrow s(y) - \text{single}(s(x)) \} 
• where: \text{single}(n) = \{n\}
 \text{single}(N) = \emptyset \text{ (otherwise)}
```

- good one: $es(p_{\neq}) = \{ fix(y), fix(x) \}$
- but also: es(p_≠) = { any(y), any(x)}

Taking Advantage from Event Sets

 Base decision of propagators to re-propagate on event sets rather than on modified variables

$$DP := \{ q \in P \mid \text{events}(s,s') \cap \text{es}(q) \neq \emptyset \};$$

Note that the MV set is not needed any more.

More Optimizations

Priorities

- Choose propagator
 - according to cost: cheapest first
 - according to expected impact
 - general: first-in first-out (queue)

Propagator Rewriting

- Another observation: propagator for max(x,y)=z and values for x are smaller than for y
- Replace by propagator for y=z

Summary: Optimizing Propagation

- Fixpoint knowledge avoids useless execution
 - idempotence, subsumption, events
 - knowledge provided by propagator
- More details on optimizing propagation and propagation in systems
 - Finite Domain Constraint Programming Systems, Christian Schulte, Mats Carlsson.

In: Francesca Rossi, Peter van Beek, Toby Walsh, editors, *Handbook of Constraint Programming*, Foundations of Artificial Intelligence, pages 495-526. Elsevier Science Publishers, 2006.