

Invariants

Motivation

Example

Consider the goal formula

$$AonB \wedge BonC$$

regressed with operator

$$\langle AonC \wedge Aclear \wedge Bclear, AonB \wedge \neg Bclear \wedge Cclear \rangle$$

giving new goal

$$AonC \wedge Aclear \wedge Bclear \wedge BonC.$$

It is intuitively clear that no state satisfying this formula is reachable by any plan from a legal blocks world state.

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- Goal formulae and formulae obtained by regression from them often represent some states that are not reachable from the initial state.
- If **none of the states** is reachable from the initial state because **there are no plans** reaching the formula.
- We would like to have **reachable states** only, if possible.
- Same problem shows up in **satisfiability planning**: **partial valuations** considered by satisfiability algorithms may represent unreachable states, and this may result in unnecessary search.

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Goal: Restriction to states that are reachable.

Problem: Testing reachability is computationally as complex as testing whether a plan exists.

Solution: Use an **approximate** notion of reachability.

Implementation: Compute in polynomial time **formulae** that characterize **a superset** of the reachable states.

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Invariants: definition

Definition

A formula ϕ is an **invariant** of $\langle A, I, O, G \rangle$ if

- 1 $I \models \phi$, and
- 2 for every $o \in O$ and state s such that $s \models \phi$ and s is reachable from I , also $app_o(s) \models \phi$.

Stated differently...

ϕ is true in every state that is reachable from I by some sequence of operators.

Example

If $l \in D_i^{max}$ for all $i \geq 1$ then l is an invariant.

Hence our algorithm for computing the sets D_i^{max} is a method for identifying **a restricted class of invariants**.

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Invariants: the strongest invariant

Definition

An invariant ϕ is **the strongest invariant** of $\langle A, I, O, G \rangle$ if for any invariant ψ , $\phi \models \psi$.

The strongest invariant **exactly characterizes** the set of all states that are reachable from the initial state:

For all states s , $s \models \phi$ if and only if s is reachable.

Remark

There are infinitely many strongest invariants, but they are all logically equivalent. (If ϕ is a strongest invariant, then so is $\phi \vee \phi \dots$)

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Example: the strongest invariant for blocks world

The strongest invariant for the blocks world

Let X be the set of blocks, for example $X = \{A, B, C, D\}$.
The conjunction of the following formulae is the **strongest invariant** for the set of all states for the blocks X .

$\text{clear}(x) \leftrightarrow \forall y \in X \setminus \{x\}. \neg \text{on}(y, x)$ for all x

$\text{ontable}(x) \leftrightarrow \forall y \in X \setminus \{x\}. \neg \text{on}(x, y)$ for all x

$\neg \text{on}(x, y) \vee \neg \text{on}(x, z)$ for all x, y, z such that $y \neq z$

$\neg \text{on}(y, x) \vee \neg \text{on}(z, x)$ for all x, y, z such that $y \neq z$

$\neg(\text{on}(x_1, x_2) \wedge \text{on}(x_2, x_3) \wedge \dots \wedge \text{on}(x_{n-1}, x_n) \wedge \text{on}(x_n, x_1))$

for every $n \geq 1$ and $\{x_1, \dots, x_n\} \subseteq X$

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Invariants: connection to plan existence

Theorem

Let ϕ be the strongest invariant for $\langle A, I, O, G \rangle$. Then $\langle A, I, O, G \rangle$ has a plan if and only if $G \wedge \phi$ is satisfiable.

Proof.

Very easy! □

Theorem

Computing the strongest invariant ϕ is PSPACE-hard.

Proof.

By reduction from the **plan existence problem**.

Fact: Testing plan existence is PSPACE-hard for $\langle A, I, O, G \rangle$ even when $G = q$ for a state variable $q \in A$. (We'll show this in two weeks!)

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Invariants: connection to plan existence

Proof continues..

Let $o = \langle q, a_1 \wedge \dots \wedge a_n \rangle$ with $A = \{a_1, \dots, a_n, q\}$.

For $\langle A, I, O, q \rangle$ a plan exists

iff for $\langle A, I, O \cup \{o\}, q \rangle$ a plan exists

iff for $\langle A, I, O \cup \{o\}, q \wedge a_1 \wedge \dots \wedge a_n \rangle$ a plan exists.

Testing satisfiability of $\phi \wedge q \wedge a_1 \wedge \dots \wedge a_n$ can be done in polynomial time: replace every state variable in the strongest invariant ϕ by \top and simplify, getting \top or \perp .

So, if we had a polynomial-time algorithm for computing the strongest invariant ϕ , we could test plan existence in polynomial time.

Hence plan existence is polynomial-time reducible to computing the strongest invariant.

Since the former is PSPACE-hard also the latter is PSPACE-hard. □

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Testing satisfiability of $\phi \wedge q \wedge a_1 \wedge \dots \wedge a_n$ can be done in polynomial time: replace every state variable in the strongest invariant ϕ by \top and simplify, getting \top or \perp .

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Computation of invariants: informally

Similar to distance estimation with D_i^{max} : compute sets C_i of n -literal clauses characterizing (giving an upper bound!) the states that are reachable in i steps.

Example

$$C_0 = \{a, \neg b, c\} \sim \{101\}$$

$$C_1 = \{a \vee b, \neg a \vee \neg b, c\} \sim \{101, 011\} \quad a, \neg b \text{ falsified}$$

$$C_2 = \{\neg a \vee \neg b, c\} \sim \{001, 011, 101\} \quad a \vee b \text{ falsified}$$

$$C_3 = \{\neg a \vee \neg b, c \vee a\} \sim \{001, 011, 100, 101\} \quad c \text{ falsified}$$

$$C_4 = \{\neg a \vee \neg b\} \sim \{000, 001, 010, 011, 100, 101\} \quad c \vee a \text{ falsified}$$

$$C_5 = \{\neg a \vee \neg b\} \sim \{000, 001, 010, 011, 100, 101\}$$

$$C_i = C_5 \text{ for all } i > 5$$

$\neg a \vee \neg b$ is the only invariant found.

Computation of invariants: informally

- 1 Start with all 1-literal clauses that are true in the initial state.
- 2 Repeatedly test every operator vs. every clause, whether the clause can be shown to be true after applying the operator:
 - 1 One of the literals in the clause is necessarily true: **retain**.
 - 2 Otherwise, if the clause is too long: **forget it**.
 - 3 Otherwise, replace the clause by **new clauses** obtained by adding literals that are now true.
- 3 When all clauses remain, stop: they are invariants.

Computation of invariants

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Let $C_0 = \{Aclear, \neg Bclear, AonB, \neg BonA, \neg AonT, BonT\}$
and $o = \langle Aclear \wedge AonB, Bclear \wedge \neg AonB \wedge AonT \rangle$.

- 1 $C_0 \cup \{Aclear \wedge AonB\}$ is satisfiable: o is applicable.
- 2 The 1-literal clauses $\neg Bclear, AonB$ and $\neg AonT$ become false when o is applied.
- 3 They are not thrown away, like we did when computing D_i^{max} . They are replaced by **weaker** clauses.
- 4 Literals true after applying o in state s such that $s \models C$: $Aclear, Bclear, \neg AonB, \neg BonA, AonT, BonT$
- 5 2-literal clauses that are **weaker than** $AonB$ and **now true** are $AonB \vee Aclear, AonB \vee Bclear, AonB \vee \neg AonB, AonB \vee \neg BonA, AonB \vee AonT, AonB \vee BonT$.

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Let $C_0 = \{Aclear, \neg Bclear, AonB, \neg BonA, \neg AonT, BonT\}$
and $o = \langle Aclear \wedge AonB, Bclear \wedge \neg AonB \wedge AonT \rangle$.

- 1 $C_0 \cup \{Aclear \wedge AonB\}$ is satisfiable: o is applicable.
- 2 The 1-literal clauses $\neg Bclear, AonB$ and $\neg AonT$ become false when o is applied.
- 3 They are not thrown away, like we did when computing D_i^{max} . They are replaced by **weaker** clauses.
- 4 Literals true after applying o in state s such that $s \models C$: $Aclear, Bclear, \neg AonB, \neg BonA, AonT, BonT$
- 5 2-literal clauses that are **weaker than** $AonB$ and **now true** are $AonB \vee Aclear, AonB \vee Bclear, AonB \vee \neg AonB, AonB \vee \neg BonA, AonB \vee AonT, AonB \vee BonT$.

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Let $C_0 = \{Aclear, \neg Bclear, AonB, \neg BonA, \neg AonT, BonT\}$
and $o = \langle Aclear \wedge AonB, Bclear \wedge \neg AonB \wedge AonT \rangle$.

- 1 $C_0 \cup \{Aclear \wedge AonB\}$ is satisfiable: o is applicable.
- 2 The 1-literal clauses $\neg Bclear$, $AonB$ and $\neg AonT$ become false when o is applied.
- 3 They are not thrown away, like we did when computing D_i^{max} . They are replaced by **weaker** clauses.
- 4 Literals true after applying o in state s such that $s \models C$:
 $Aclear, Bclear, \neg AonB, \neg BonA, AonT, BonT$
- 5 2-literal clauses that are **weaker than** $AonB$ and **now true** are $AonB \vee Aclear, AonB \vee Bclear, AonB \vee \neg AonB, AonB \vee \neg BonA, AonB \vee AonT, AonB \vee BonT$.

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Let $C_0 = \{Aclear, \neg Bclear, AonB, \neg BonA, \neg AonT, BonT\}$
and $o = \langle Aclear \wedge AonB, Bclear \wedge \neg AonB \wedge AonT \rangle$.

- 1 $C_0 \cup \{Aclear \wedge AonB\}$ is satisfiable: o is applicable.
- 2 The 1-literal clauses $\neg Bclear, AonB$ and $\neg AonT$ become false when o is applied.
- 3 They are not thrown away, like we did when computing D_i^{max} . They are replaced by **weaker** clauses.
- 4 Literals true after applying o in state s such that $s \models C$: $Aclear, Bclear, \neg AonB, \neg BonA, AonT, BonT$
- 5 2-literal clauses that are **weaker than** $AonB$ and **now true** are $AonB \vee Aclear, AonB \vee Bclear, AonB \vee \neg AonB, AonB \vee \neg BonA, AonB \vee AonT, AonB \vee BonT$.

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Example continues..

- 7 For $\neg Bclear$ and $\neg AonT$ we respectively get $\neg Bclear \vee Aclear$, $\neg Bclear \vee Bclear$, $\neg Bclear \vee \neg AonB$, $\neg Bclear \vee \neg BonA$, $\neg Bclear \vee AonT$, $\neg Bclear \vee BonT$ **and** $\neg AonT \vee Aclear$, $\neg AonT \vee Bclear$, $\neg AonT \vee \neg AonB$, $\neg AonT \vee \neg BonA$, $\neg AonT \vee AonT$, $\neg AonT \vee BonT$.
- 8 By eliminating logically equivalent ones, tautologies, and those that follow from those in C_0 not falsified we get $C_1 = \{Aclear, \neg BonA, BonT, AonB \vee Bclear, AonB \vee AonT, \neg Bclear \vee \neg AonB, \neg Bclear \vee AonT, \neg AonT \vee Bclear, \neg AonT \vee \neg AonB\}$ for distance 1 states.
- 9 The precondition of $\langle Bclear \wedge BonT \wedge Aclear, BonA \wedge \neg Aclear \wedge \neg BonT \rangle$ is satisfiable with C_1 , and the set C_2 contains all invariants for 2 blocks.

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Let $C_i = \{\neg \text{AinRome} \vee \neg \text{AinNYC}, \neg \text{AinParis} \vee \neg \text{AinNYC}, \neg \text{AinParis} \vee \neg \text{AinNYC}\}$,
 $o = \langle \text{AinRome}, \text{AinParis} \wedge \neg \text{AinRome} \rangle$.

- 1 Does o preserve truth of $\neg \text{AinParis} \vee \neg \text{AinNYC}$?
- 2 Because o makes $\neg \text{AinParis}$ false, we must show that $\neg \text{AinNYC}$ is true after applying o .
- 3 But $\neg \text{AinNYC}$ is not even mentioned in o !
- 4 However, since AinRome is the precondition of o and $\neg \text{AinRome} \vee \neg \text{AinNYC}$ was true before applying o , we can infer that $\neg \text{AinNYC}$ was true before applying o .
- 5 Since o does not make $\neg \text{AinNYC}$ false, it is true also after applying o , and then so is $\neg \text{AinParis} \vee \neg \text{AinNYC}$.

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Let $C_i = \{\neg \text{AinRome} \vee \neg \text{AinNYC}, \neg \text{AinParis} \vee \neg \text{AinNYC}, \neg \text{AinParis} \vee \neg \text{AinNYC}\}$,
 $o = \langle \text{AinRome}, \text{AinParis} \wedge \neg \text{AinRome} \rangle$.

- 1 Does o preserve truth of $\neg \text{AinParis} \vee \neg \text{AinNYC}$?
- 2 Because o makes $\neg \text{AinParis}$ false, we must show that $\neg \text{AinNYC}$ is true after applying o .
- 3 But $\neg \text{AinNYC}$ is not even mentioned in o !
- 4 However, since AinRome is the precondition of o and $\neg \text{AinRome} \vee \neg \text{AinNYC}$ was true before applying o , we can infer that $\neg \text{AinNYC}$ was true before applying o .
- 5 Since o does not make $\neg \text{AinNYC}$ false, it is true also after applying o , and then so is $\neg \text{AinParis} \vee \neg \text{AinNYC}$.

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Let $C_i = \{\neg \text{AinRome} \vee \neg \text{AinNYC}, \neg \text{AinParis} \vee \neg \text{AinNYC}, \neg \text{AinParis} \vee \neg \text{AinNYC}\}$,
 $o = \langle \text{AinRome}, \text{AinParis} \wedge \neg \text{AinRome} \rangle$.

- 1 Does o preserve truth of $\neg \text{AinParis} \vee \neg \text{AinNYC}$?
- 2 Because o makes $\neg \text{AinParis}$ false, we must show that $\neg \text{AinNYC}$ is true after applying o .
- 3 But $\neg \text{AinNYC}$ is not even mentioned in o !
- 4 However, since AinRome is the precondition of o and $\neg \text{AinRome} \vee \neg \text{AinNYC}$ was true before applying o , we can infer that $\neg \text{AinNYC}$ was true before applying o .
- 5 Since o does not make $\neg \text{AinNYC}$ false, it is true also after applying o , and then so is $\neg \text{AinParis} \vee \neg \text{AinNYC}$.

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Let $C_i = \{\neg \text{AinRome} \vee \neg \text{AinNYC}, \neg \text{AinParis} \vee \neg \text{AinNYC}, \neg \text{AinParis} \vee \neg \text{AinNYC}\}$,
 $o = \langle \text{AinRome}, \text{AinParis} \wedge \neg \text{AinRome} \rangle$.

- 1 Does o preserve truth of $\neg \text{AinParis} \vee \neg \text{AinNYC}$?
- 2 Because o makes $\neg \text{AinParis}$ false, we must show that $\neg \text{AinNYC}$ is true after applying o .
- 3 But $\neg \text{AinNYC}$ is not even mentioned in o !
- 4 However, since AinRome is the precondition of o and $\neg \text{AinRome} \vee \neg \text{AinNYC}$ was true before applying o , we can infer that $\neg \text{AinNYC}$ was true before applying o .
- 5 Since o does not make $\neg \text{AinNYC}$ false, it is true also after applying o , and then so is $\neg \text{AinParis} \vee \neg \text{AinNYC}$.

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Let $C_i = \{\neg \text{AinRome} \vee \neg \text{AinNYC}, \neg \text{AinParis} \vee \neg \text{AinNYC}, \neg \text{AinParis} \vee \neg \text{AinNYC}\}$,
 $o = \langle \text{AinRome}, \text{AinParis} \wedge \neg \text{AinRome} \rangle$.

- 1 Does o preserve truth of $\neg \text{AinParis} \vee \neg \text{AinNYC}$?
- 2 Because o makes $\neg \text{AinParis}$ false, we must show that $\neg \text{AinNYC}$ is true after applying o .
- 3 But $\neg \text{AinNYC}$ is not even mentioned in o !
- 4 However, since AinRome is the precondition of o and $\neg \text{AinRome} \vee \neg \text{AinNYC}$ was true before applying o , we can infer that $\neg \text{AinNYC}$ was true before applying o .
- 5 Since o does not make $\neg \text{AinNYC}$ false, it is true also after applying o , and then so is $\neg \text{AinParis} \vee \neg \text{AinNYC}$.

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Let $C_i = \{\neg \text{AinRome} \vee \neg \text{AinNYC}, \neg \text{AinParis} \vee \neg \text{AinNYC}, \neg \text{AinParis} \vee \neg \text{AinNYC}\}$,
 $o = \langle \text{AinRome}, \text{AinParis} \wedge \neg \text{AinRome} \rangle$.

- 1 Does o preserve truth of $\neg \text{AinParis} \vee \neg \text{AinNYC}$?
- 2 Because o makes $\neg \text{AinParis}$ false, we must show that $\neg \text{AinNYC}$ is true after applying o .
- 3 But $\neg \text{AinNYC}$ is not even mentioned in o !
- 4 However, since AinRome is the precondition of o and $\neg \text{AinRome} \vee \neg \text{AinNYC}$ was true before applying o , we can infer that $\neg \text{AinNYC}$ was true before applying o .
- 5 Since o does not make $\neg \text{AinNYC}$ false, it is true also after applying o , and then so is $\neg \text{AinParis} \vee \neg \text{AinNYC}$.

Computation of invariants: procedure *preserved*

Test whether a clause remains true when operator is applied

```
PROCEDURE preserved( $l_1 \vee \dots \vee l_n, C, o$ );  
 $\langle c, e \rangle := o$ ;  
FOR EACH  $l \in \{l_1, \dots, l_n\}$  DO  
  IF  $C \cup \{EPC_{\bar{l}}(o)\}$  is unsatisfiable THEN GOTO OK;  
  FOR EACH  $l' \in \{l_1, \dots, l_n\} \setminus \{l\}$  DO  
    IF  $C \cup \{EPC_{\bar{l}}(o)\} \models EPC_{l'}(e)$  THEN GOTO OK;  
    IF  $C \cup \{EPC_{\bar{l}}(o)\} \models l' \wedge \neg EPC_{\bar{l}}(e)$  THEN GOTO OK;  
  END DO  
  RETURN false;  
OK:  
END DO  
RETURN true;
```

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Computation of invariants: function *preserved*

Let $C = \{c \vee b\}$.

- 1 **preserved($a \vee b, C, \langle \neg c, c \wedge d \rangle$) returns *true***
- 2 preserved($a \vee b, C, \langle \neg c, \neg a \wedge b \rangle$) returns *true*
- 3 preserved($a \vee b, C, \langle b, \neg a \rangle$) returns *true*
- 4 preserved($a \vee b, C, \langle \neg c, \neg a \rangle$) returns *true*
- 5 preserved($a \vee b, C, \langle c, \neg a \rangle$) returns *false*

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- 3 $\text{preserved}(a \vee b, C, \langle b, \neg a \rangle)$ returns *true*
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- 4 $\text{preserved}(a \vee b, C, \langle \neg c, \neg a \rangle)$ returns *true*
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Lemma

Let C be a set of clauses, $\phi = l_1 \vee \dots \vee l_n$ a clause, and o an operator.

If $\text{preserved}(\phi, C, o)$ returns true, then $\text{app}_o(s) \models \phi$ for every state s such that $s \models C$ and $\text{app}_o(s)$ is defined.

Computation of invariants: function *preserved*

Why is *preserved* incomplete?

Example

Let $o = \langle a, \neg b \wedge (c \triangleright d) \wedge (\neg c \triangleright e) \rangle$.

$\text{preserved}(b \vee d \vee e, \emptyset, o)$ returns **false** because it cannot prove for **any literal** in $b \vee d \vee e$ that it is true after application of o . However, $d \vee e$ is true after applying o , and hence $b \vee d \vee e$ will be true as well.

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Outline

- 1 C = the set of 1-literal clauses that are true in the initial state.
- 2 For each operator o and clause $l_1 \vee \dots \vee l_m \in C$ test if $l_1 \vee \dots \vee l_m$ remains true when o is applied.
- 3 If not, remove $l_1 \vee \dots \vee l_m$, and if $m < n$ add clauses $l_1 \vee \dots \vee l_m \vee a$ and $l_1 \vee \dots \vee l_m \vee \neg a$ for every $a \in A$.
- 4 Repeat from step 2 if C has changed.
- 5 Otherwise every clause in C is an invariant.

The number of iterations is $\mathcal{O}(m^n)$ which is polynomial in the number of state variables $m = |A|$ for any fixed n .

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```
PROCEDURE invariants( $A, I, O, n$ );  
 $C := \{a \in A \mid I \models a\} \cup \{\neg a \mid a \in A, I \not\models a\}$ ;  
REPEAT  
   $C' := C$ ;  
  FOR EACH  $l_1 \vee \dots \vee l_m \in C$  AND  $o \in O$   
    such that  $\text{preserved}(l_1 \vee \dots \vee l_m, C', o) = \text{false}$  DO  
     $C := C \setminus \{l_1 \vee \dots \vee l_m\}$ ;  
    IF  $m < n$  THEN  
       $C := C \cup \bigcup_{a \in A} \{l_1 \vee \dots \vee l_m \vee a, l_1 \vee \dots \vee l_m \vee \neg a\}$ ;  
    END FOR  
UNTIL  $C = C'$ ;  
RETURN  $C$ ;
```

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Theorem

The procedure $\text{invariants}(A, I, O, n)$ returns a set C of clauses with at most n literals so that for any sequence o_1, \dots, o_m of operators in O $\text{app}_{o_1; \dots; o_m}(I) \models C$.

Proof.

Let C_0 be the value first assigned to the variable C and C_1, C_2, \dots the values of C in the end of each iteration.

Induction hypothesis: for every $\{o_1, \dots, o_i\} \subseteq O$ and $\phi \in C_i$, $\text{app}_{o_1; \dots; o_i}(I) \models \phi$.

Base case $i = 0$: $\text{app}_\epsilon(I)$ for the empty sequence is by definition I itself, and by construction C_0 consists of only formulae that are true in the initial state.

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Proof continues..

Inductive case $i \geq 1$: Take any $\{o_1, \dots, o_i\} \subseteq O$ and $\phi \in C_i$.

A Consider the case $\phi \in C_{i-1}$. By induction hypothesis $app_{o_1; \dots; o_{i-1}}(I) \models \phi$. Since $\phi \in C_i$ $preserved(\phi, C_{i-1}, o)$ returns *true*. Hence by the Lemma $app_{o_1; \dots; o_i}(I) \models \phi$.

B Consider the case $\phi \notin C_{i-1}$.

- 1 As $\phi \notin C_{i-1}$ there is $\phi' \in C_{i-1}$ with $\phi = \phi' \vee l'_1 \vee \dots \vee l'_m$ for some l'_1, \dots, l'_m and $preserved(\phi', C_{i-1}, o')$ returns *false* for some $o' \in O$. Hence $\phi' \not\models \phi$.
- 2 As $\phi' \in C_{i-1}$ by induction hypothesis $app_{o_1; \dots; o_{i-1}}(I) \models \phi'$.
- 3 Since $\phi' \not\models \phi$ also $app_{o_1; \dots; o_{i-1}}(I) \not\models \phi$.
- 4 Since $preserved(\phi, C_i, o)$ returns *true* $app_{o_1; \dots; o_i}(I) \models \phi$ by the Lemma.



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Why is the strongest invariant not always found?

- 1 Practical implementations of the algorithm use **polynomial time approximations** of the tests for satisfiability and \models .
- 2 The function *preserved* is incomplete for operators in general (but complete for STRIPS operators.) Making it complete makes it NP-hard.
- 3 The strongest invariant may require **arbitrarily long clauses**, so the restriction to clauses of any **fixed length** makes it impossible to represent it.

Example

The acyclicity of the **on** relation in the blocks world needs clauses of length n when there are n blocks.

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Computation of invariants

Example

Initial state: $I \models a \wedge \neg b \wedge \neg c$

Operators: $o_1 = \langle a, \neg a \wedge b \rangle,$

$o_2 = \langle b, \neg b \wedge c \rangle,$

$o_3 = \langle c, \neg c \wedge a \rangle$

Computation: Find invariants with at most 2 literals:

$$C_0 = \{a, \neg b, \neg c\}$$

$$C_1 = \{\neg c, a \vee b, \neg b \vee \neg a\}$$

$$C_2 = \{\neg b \vee \neg a, \neg c \vee \neg a, \neg c \vee \neg b\}$$

$$C_3 = \{\neg b \vee \neg a, \neg c \vee \neg a, \neg c \vee \neg b\}$$

$$C_j = C_2 \text{ for all } j \geq 2$$

Invariants in satisfiability planning

Invariants in satisfiability planning

For every invariant $l_1 \vee \dots \vee l_n$ add the clauses

$$l_1^t \vee \dots \vee l_n^t$$

for all time points t .

Notice that the above formulae **logical consequences** of Φ_i^{seq} and Φ_i^{par} , so the invariants do not change the set of **valuations** of these formulae.

Invariants are critical for the efficiency of satisfiability planning on many types of problems.

Invariants in backward search

Motivating example

Example

Regression of $\text{in}(A, \text{Freiburg})$ by

$\langle \text{in}(A, \text{Strassburg}), \neg \text{in}(A, \text{Strassburg}) \wedge \text{in}(A, \text{Paris}) \rangle$

gives $\text{in}(A, \text{Freiburg}) \wedge \text{in}(A, \text{Strassburg})$

No state satisfying $\text{in}(A, \text{Freiburg}) \wedge \text{in}(A, \text{Strassburg})$ makes sense if A denotes some usual physical object.

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Regression

Summary

Invariants in backward search

Motivating example

Problem: Regression produces sets T of states such that

- 1 some states in T are not reachable from I , or
- 2 none of the states in T are reachable from I .

The first is not always a serious problem (but may worsen the quality of distance estimates, for example.)

Solution: Use invariants to avoid formulae that do not represent any reachable states.

- 1 Compute invariant ϕ .
- 2 Do only regression steps such that $\text{regr}_o(\psi) \wedge \phi$ is satisfiable.

Invariants in backward search

Motivating example

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Summary

- Invariants are needed for making **backward search** and **satisfiability planning** more efficient.
- We gave an algorithm for computing a class of invariants.
 - Start with 1-literal clauses true in the initial state.
 - Repeatedly weaken clauses that could not be shown to be invariants.
 - Stop when all clauses are guaranteed to be invariants.
- The algorithm runs in polynomial time if the satisfiability and logical consequence tests are approximated by a polynomial time algorithm.

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