

Principles of AI Planning

Invariants

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Example

Consider the goal formula

$$AonB \wedge \textcolor{red}{BonC}$$

regressed with operator

$$\langle \textcolor{red}{AonC} \wedge Aclear \wedge Bclear, AonB \wedge \neg Bclear \wedge Cclear \rangle$$

resulting in the new goal

$$\textcolor{red}{AonC} \wedge Aclear \wedge Bclear \wedge \textcolor{red}{BonC}.$$

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 regressing them often represent some states that are not reachable from the initial state.
- If **none of the states** is reachable from the initial state, **there are no plans** reaching the formula.
- We would like to have **reachable states** only, if possible.
- The 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 $s \models \phi$ for every state s reachable from I .

Example

The formula $\neg(AonB \wedge AonC)$ is an invariant in a blocks world task.

Remark

Invariants are usually proved inductively:

- Prove that ϕ is true in the initial state.
- Prove that operator application preserves ϕ .

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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 for any given planning task, 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 .

For all $x \in X$: $\text{clear}(x) \leftrightarrow \bigwedge_{y \in X} \neg \text{on}(y, x)$

For all $x \in X$: $\text{ontable}(x) \leftrightarrow \bigwedge_{y \in X} \neg \text{on}(x, y)$

For all $x, y, z \in X$ with $y \neq z$: $\neg \text{on}(x, y) \vee \neg \text{on}(x, z)$

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For all $n \geq 1$ and $x_1, \dots, x_n \in X$:

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

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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.
Even deciding whether or not \top is the strongest invariant is already PSPACE-hard.*

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Invariants: connection to plan existence

Proof.

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

Fact: Testing plan existence for $\langle A, I, O, G \rangle$ is PSPACE-hard.
(We'll show this later this month!)

Let $a' \notin A$ be a new state variable. Then a plan exists for $\mathcal{T} = \langle A, I, O, G \rangle$ iff \top is the strongest invariant of the planning task $\mathcal{T}' = \langle A \cup \{a'\}, I \cup \{a' \mapsto 0\}, O \cup O', G \rangle$, where

$$O' = \{ \langle G, a' \wedge \bigwedge_{a \in A} a \rangle \} \\ \cup \{ \langle a', \neg a \rangle \mid a \in A \cup \{a'\} \}.$$

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Proof continues...

(\Rightarrow): If a plan exists for \mathcal{T} , then the same plan is applicable in \mathcal{T}' . We can thus reach a state satisfying G in \mathcal{T}' .

From this state, we can reach *any* state s by first applying $\langle G, a' \wedge \bigwedge_{a \in A} a \rangle$ and then applying the operators $\langle a', \neg a \rangle$ for each variable a with $s(a) = 0$. (If $s(a') = 0$, the corresponding operator must be applied last.)

If *all* states are reachable in \mathcal{T}' , then \top is the strongest invariant for \mathcal{T}' .

(\Leftarrow) (by contraposition): If \mathcal{T} is not solvable, then no state satisfying G is reachable in \mathcal{T} . In that case, no state satisfying G is reachable in \mathcal{T}' , and thus a' cannot be made true in \mathcal{T}' . Thus, $\neg a'$ is an invariant in \mathcal{T}' which is stronger than \top , so \top is not the strongest invariant in \mathcal{T}' . \square

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Computation of invariants: informally

Compute sets C_i of *n*-literal clauses characterizing (giving an upper bound!) the states that are reachable in i steps.

Example

$$\begin{aligned}C_0 &= \{a, \neg b, c\} && \sim \{101\} \\C_1 &= \{a \vee b, \neg a \vee \neg b, c\} && \sim \{101, 011\} \\C_2 &= \{\neg a \vee \neg b, c\} && \sim \{001, 011, 101\} \\C_3 &= \{\neg a \vee \neg b, c \vee a\} && \sim \{001, 011, 100, 101\} \\C_4 &= \{\neg a \vee \neg b\} && \sim \{000, 001, 010, 011, 100, 101\} \\C_5 &= \{\neg a \vee \neg b\} && \sim \{000, 001, 010, 011, 100, 101\} \\C_i &= C_5 \text{ for all } i > 5\end{aligned}$$

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

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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 to check 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 are retained, stop: they are invariants.

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- 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, though: they are replaced by **weaker** clauses.
- 4 Literals true after applying o in state s such that $s \models C_0$: $Aclear$, $Bclear$, $\neg AonB$, $\neg BonA$, $AonT$, $BonT$.
- 5 2-literal clauses that are **weaker than** $\neg Bclear$ and **now true** are $\neg Bclear \vee Aclear$, $\neg Bclear \vee Bclear$, $\neg Bclear \vee \neg AonB$, $\neg Bclear \vee \neg BonA$, $\neg Bclear \vee AonT$, and $\neg Bclear \vee BonT$.

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- ➍ Literals true after applying o in state s such that $s \models C_0$: $Aclear$, $Bclear$, $\neg AonB$, $\neg BonA$, $AonT$, $BonT$.
- ➎ 2-literal clauses that are **weaker than** $\neg Bclear$ and **now true** are $\neg Bclear \vee Aclear$, $\neg Bclear \vee Bclear$, $\neg Bclear \vee \neg AonB$, $\neg Bclear \vee \neg BonA$, $\neg Bclear \vee AonT$, and $\neg Bclear \vee BonT$.

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- ② The 1-literal clauses $\neg Bclear$, $AonB$ and $\neg AonT$ become false when o is applied.
- ③ They are not thrown away, though: they are replaced by **weaker** clauses.
- ④ Literals true after applying o in state s such that $s \models C_0$: $Aclear$, $Bclear$, $\neg AonB$, $\neg BonA$, $AonT$, $BonT$.
- ⑤ 2-literal clauses that are **weaker than** $\neg Bclear$ and **now true** are $\neg Bclear \vee Aclear$, $\neg Bclear \vee Bclear$, $\neg Bclear \vee \neg AonB$, $\neg Bclear \vee \neg BonA$, $\neg Bclear \vee AonT$, and $\neg Bclear \vee BonT$.

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Example (continues. . .)

- ⑥ Similar 2-literal clauses are obtained from $AonB$ and from $\neg AonT$.
- ⑦ By eliminating logically equivalent ones, tautologies, and clauses that follow from those in C_0 not falsified we get $C_1 = \{Aclear, \neg BonA, BonT, \neg Bclear \vee \neg AonB, \neg Bclear \vee AonT, AonB \vee Bclear, AonB \vee AonT, \neg AonT \vee Bclear, \neg AonT \vee \neg AonB\}$ for distance 1 states.
- ⑧ Some clauses in C_1 can be refined further by checking other operators whose preconditions are consistent with C_1 . With a bit more computation, C_i settles to a set containing all invariants for two blocks.

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

- ❶ Does o preserve truth of $\neg \text{AinParis} \vee \neg \text{AinNYC}$?
- ❷ Because o makes $\neg \text{AinParis}$ false, we must show that $\neg \text{AinNYC}$ is true after applying o .
- ❸ But $\neg \text{AinNYC}$ is **not even mentioned** in o !
- ❹ 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 .
- ❺ 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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Computation of invariants: function *preserved*

Test whether a clause remains true when operator is applied

Test if an operator preserves a clause

```
def preserved( $l_1 \vee \dots \vee l_n, C, o$ ):  
    for each  $l \in \{l_1, \dots, l_n\}$ :  
        if not preserved-literal( $C, o, \{l_1, \dots, l_n\} \setminus \{l\}, l$ ):  
            return false  
    return true
```

Test if an operator preserves a literal

```
def preserved-literal( $C, o, L', l$ ):  
     $\langle c, e \rangle := o$   
     $C_{\bar{l}} := C \cup \{c\} \cup \{EPC_{\bar{l}}(e)\}$   
    return  $C_{\bar{l}}$  is unsatisfiable  
    or  $C_{\bar{l}} \models EPC_{l'}(e)$  for some  $l' \in L'$   
    or  $C_{\bar{l}} \models l' \wedge \neg EPC_{\bar{l}'}(e)$  for some  $l' \in L'$ 
```

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Computation of invariants: function *preserved*

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

- ➊ **`preserved($a \vee b$, C , $\langle \neg c, c \wedge d \rangle$)` returns *true***
- ➋ `preserved($a \vee b$, C , $\langle \neg c, \neg a \wedge b \rangle$)` returns *true*
- ➌ `preserved($a \vee b$, C , $\langle b, \neg a \rangle$)` returns *true*
- ➍ `preserved($a \vee b$, C , $\langle \neg c, \neg a \rangle$)` returns *true*
- ➎ `preserved($a \vee b$, C , $\langle c, \neg a \rangle$)` returns *false*

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Computation of invariants: function *preserved*

Correctness

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 $s \models C \cup \{\phi\}$ and $\text{app}_o(s)$ is defined.

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Why is *preserved* incomplete?

Example (incompleteness)

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 the preserved-literal check for $l = b$ fails:

- Operator o can make b false.
- It is **not guaranteed** that d is true in the resulting state.
- It is **not guaranteed** that e is true in the resulting state.

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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Computation of invariants: the main procedure

Outline

- 1 C = the set of 1-literal clauses that are true in the initial state.
- 2 For each operator o and clause $\phi \in C$, test if ϕ remains true when o is applied.
- 3 If not, remove ϕ , and if the number of literals in ϕ is less than n , add clauses $\phi \vee l$ for each literal l which is guaranteed to be true after applying o .
- 4 Remove all dominated invariants.
- 5 Repeat from step 2 if C has changed in the previous two steps.
- 6 Otherwise every clause in C is an invariant.

For any **fixed limit** n on the size of the clauses, the number of iterations is $\mathcal{O}(m^n)$ (where $m = |A|$ is the number of state variables) and hence polynomial.

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Computation of invariants: the main procedure

Invariant computation

```
def 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  $o = \langle c, e \rangle \in O$   
            with preserved( $l_1 \vee \dots \vee l_m, C', o$ ) = false:  
                 $C := C \setminus \{l_1 \vee \dots \vee l_m\}$   
                if  $m < n$ :  
                    for each literal  $l$ :  
                        if  $C' \cup \{c\} \models EPC_l(e) \vee (l \wedge \neg EPC_{\bar{l}}(e))$ :  
                             $C := C \cup \{l_1 \vee \dots \vee l_m \vee l\}$   
     $C := \{ \phi \in C \mid \neg \exists \phi' \in C : \phi' \models \phi \wedge \phi' \not\models \phi \}$   
until  $C = C'$   
return  $C$ 
```

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Correctness

Theorem

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

Proof.

$A \quad I \models C$:

- The initial state satisfies the initial set of 1-literal clauses.
- All modifications to the clause set only make it logically weaker (i.e., $C' \models C$ after each iteration of the main loop.)
- Thus the initial state satisfies the resulting clause set C by induction over the number of iterations.

...



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Proof continues. . .

- B** If $s \models C$ and $app_o(s)$ is defined, then $app_o(s) \models C$.
- In the last iteration of the procedure, no formula is removed from $C = C'$, and hence $preserved(\phi, C, o)$ is true for all clauses $\phi \in C$ and operators $o \in O$.
 - By the lemma, this means that $app_o(s) \models \phi$ for every state s such that $s \models C$ and $app_o(s)$ is defined.
 - Since this is true for all clauses $\phi \in C$, we get $app_o(s) \models C$ for every state s such that $s \models C$ and $app_o(s)$ is defined.

From A and B, the theorem follows by induction over the length of the operator sequence. □

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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:

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

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

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

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

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

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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 are **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.

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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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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 $regr_o(\psi) \wedge \phi$ is satisfiable.

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Invariants for problem reformulation

Mutexes

Binary clause invariants are called **mutexes** because they state that certain variable assignments cannot be simultaneously true and are hence **mutually exclusive**.

Example

The invariant $\neg A \vee \neg B \vee \neg C$ states that A , B and C are mutex.

Often, a larger **set of literals** is mutually exclusive because every pair of them forms a mutex.

Example

In blocks world, $BonA$, $ConA$, $DonA$ and $Aclear$ are mutex.

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Invariants for problem reformulation

Multi-valued state variables

If a group of n literals $G = \{l_1, \dots, l_n\}$ over N different variables $A_G = \{a_1, \dots, a_n\}$ are mutually exclusive, then the planning task can be rephrased using a single **multi-valued** (i.e., non-binary) state variable v_G with $n + 1$ possible values in place of the n variables in A_G :

- n of the possible values represent situations in which exactly one of the literals in G is true.
- The remaining value represents situations in which none of the literals in G is true.

In many cases, the reduction in the number of variables can dramatically improve performance of a planning algorithm.

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- Invariants are needed for making **backward search** and **satisfiability planning** more efficient and (in the case of mutexes) can be used for **problem reformulation**.
- 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 and the size of the invariant clauses is bounded by a constant.

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