

# Planning and Optimization

## A7. Invariants and Mutexes

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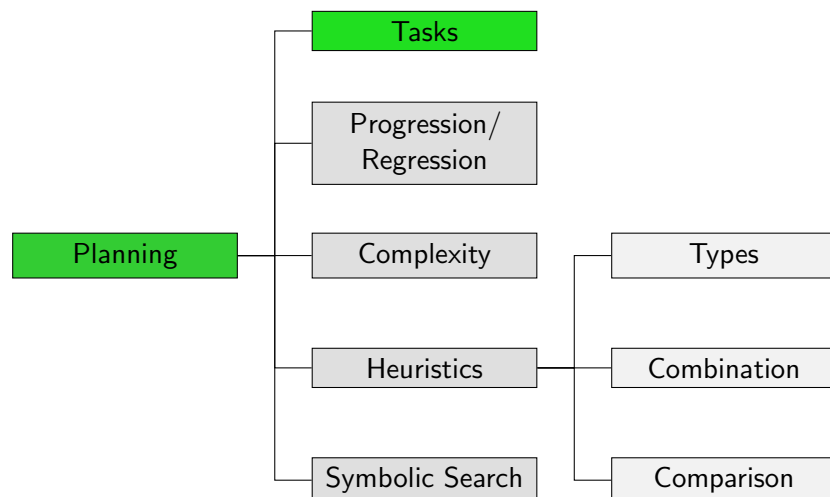
## A7.1 Invariants

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## Content of this Course



# A7.1 Invariants

## Invariants

- ▶ When we as humans reason about planning tasks, we implicitly make use of “obvious” properties of these tasks.
  - ▶ **Example:** we are never in two places at the same time
- ▶ We can represent such properties as logical formulas  $\varphi$  that are **true in all reachable states**.
  - ▶ **Example:**  $\varphi = \neg(at\text{-}uni \wedge at\text{-}home)$
- ▶ Such formulas are called **invariants** of the task.

## Invariants: Definition

### Definition (Invariant)

An **invariant** of a planning task  $\Pi$  with state variables  $V$  is a logical formula  $\varphi$  over  $V$  such that  $s \models \varphi$  for all reachable states of  $\Pi$ .

## A7.2 Computing Invariants

## Computing Invariants

How does an **automated** planner come up with invariants?

- ▶ Theoretically, testing if an arbitrary formula  $\varphi$  is an invariant is **as hard as planning** itself.
  - ↔ **proof idea:** a planning task is **unsolvable** iff the negation of its goal is an invariant
- ▶ Still, many practical invariant synthesis algorithms exist.
- ▶ To remain efficient (= polynomial-time), these algorithms only compute a **subset** of all useful invariants.
  - ↔ **sound**, but not **complete**
- ▶ Empirically, they tend to at least find the “obvious” invariants of a planning task.

## Invariant Synthesis Algorithms

Most algorithms for generating invariants are based on the **generate-test-repair** approach:

- ▶ **Generate:** Suggest some invariant candidates, e.g., by enumerating all possible formulas  $\varphi$  of a certain size.
- ▶ **Test:** Try to prove that  $\varphi$  is indeed an invariant. Usually done **inductively**:
  - ① Test that **initial state** satisfies  $\varphi$ .
  - ② Test that if  $\varphi$  is true in the current state, it remains true after applying a single operator.
- ▶ **Repair:** If invariant test fails, replace candidate  $\varphi$  by a **weaker** formula, ideally exploiting **why** the proof failed.

## Invariant Synthesis: References

We will not cover invariants in detail in this course.

Literature on invariant synthesis:

- ▶ DISCOPLAN (Gerevini & Schubert, 1998)
- ▶ TIM (Fox & Long, 1998)
- ▶ Edelkamp & Helmert's algorithm (1999)
- ▶ Bonet & Geffner's algorithm (2001)
- ▶ Rintanen's algorithm (2008)

## Exploiting Invariants

Invariants have many uses in planning:

- ▶ **Regression search:**  
Prune states that violate (are inconsistent with) invariants.
- ▶ **Planning as satisfiability:**  
Add invariants to a SAT encoding of a planning task to get tighter constraints.
- ▶ **Reformulation:**  
Derive a **more compact** state space representation (i.e., with fewer unreachable states).

We now briefly discuss the last point because it is important for **planning tasks in finite-domain representation**, introduced in the following chapter.

## A7.3 Mutexes

## Mutexes

Invariants that take the form of **binary clauses** are called **mutexes** because they express that certain variable assignments cannot be simultaneously true and are hence **mutually exclusive**.

### Example (Blocks World)

The invariant  $\neg A\text{-on-}B \vee \neg A\text{-on-}C$  states that  $A\text{-on-}B$  and  $A\text{-on-}C$  are mutex.

We say that a larger **set of literals** is mutually exclusive if every subset of two literals is a mutex.

### Example (Blocks World)

Every pair in  $\{B\text{-on-}A, C\text{-on-}A, D\text{-on-}A, A\text{-clear}\}$  is mutex.

## Encoding Mutex Groups as Finite-Domain Variables

Let  $L = \{\ell_1, \dots, \ell_n\}$  be mutually exclusive literals over  $n$  different variables  $V_L = \{v_1, \dots, v_n\}$ .

Then the planning task can be rephrased using a single **finite-domain** (i.e., non-binary) state variable  $v_L$  with  $n + 1$  possible values in place of the  $n$  variables in  $V_L$ :

- ▶  $n$  of the possible values represent situations in which **exactly one** of the literals in  $L$  is true.
- ▶ The remaining value represents situations in which **none of the literals** in  $L$  is true.
  - ▶ **Note:** If we can prove that one of the literals in  $L$  must be true in each state (i.e.,  $\ell_1 \vee \dots \vee \ell_n$  is an invariant), this additional value can be omitted.

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

## A7.4 Summary

## Summary

- ▶ **Invariants** are common properties of all reachable states, expressed as logical formulas.
- ▶ A number of algorithms for **computing invariants** exist.
- ▶ These algorithms will not find **all useful invariants** (which is too hard), but try to find some useful subset with reasonable (polynomial) computational effort.
- ▶ **Mutexes** are invariants that express that certain pairs of literals are mutually exclusive.