Planning and Optimization C4. SAT Planning: Core Idea and Sequential Encoding

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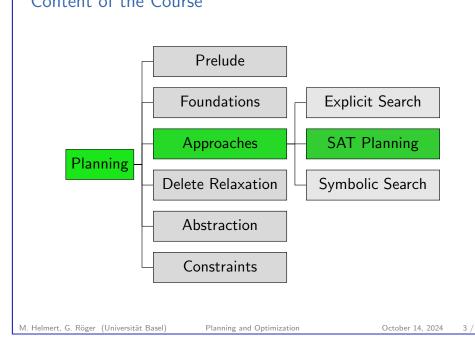
C4.5 Summary

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Content of the Course



C4. SAT Planning: Core Idea and Sequential Encoding

Introduction

C4.1 Introduction

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Introduction

SAT Solvers

- ► SAT solvers (algorithms that find satisfying assignments to CNF formulas) are one of the major success stories in solving hard combinatorial problems.
- ► Can we leverage them for classical planning?
- → SAT planning (a.k.a. planning as satisfiability)

background on SAT Solvers:

→ Foundations of Artificial Intelligence Course, Ch. E4–E5

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Solution: Iterative Deepening

- ▶ We can generate a propositional formula that tests if task Π has a plan with horizon (length bound) Tin time $O(\|\Pi\|^k \cdot T)$ (\rightsquigarrow pseudo-polynomial reduction).
- ▶ Use as building block of algorithm that probes increasing horizons (a bit like IDA*).
- ► Can be efficient if there exist plans that are not excessively long.

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Complexity Mismatch

- ► The SAT problem is NP-complete, while PLANEX is PSPACE-complete.
- \rightsquigarrow one-shot polynomial reduction from PLANEX to SAT not possible (unless NP = PSPACE)

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Introduction

SAT Planning: Main Loop

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basic SAT Planning algorithm:

```
SAT Planning
def satplan(\Pi):
     for T \in \{0, 1, 2, \dots\}:
           \varphi := \mathsf{build\_sat\_formula}(\Pi, T)
           I = \text{sat\_solver}(\varphi)
                                                > returns a model or none
           if / is not none:
                 return extract_plan(\Pi, T, I)
```

Termination criterion for unsolvable tasks?

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Formula Overview

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Formula Overview

SAT Formula: CNF?

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► SAT solvers require conjunctive normal form (CNF), i.e., formulas expressed as collection of clauses.

- ▶ We will make sure that our SAT formulas are in CNF when our input is a STRIPS task.
- ▶ We do allow fully general propositional tasks, but then the formula may need additional conversion to CNF.

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Formula Overview

SAT Formula: Variables

- ightharpoonup given propositional planning task $\Pi = \langle V, I, O, \gamma \rangle$
- ▶ given horizon $T \in \mathbb{N}_0$

Variables of the SAT Formula

- ▶ propositional variables v^i for all $v \in V$, $0 \le i \le T$ encode state after i steps of the plan
- **propositional variables** o^i for all $o \in O$, 1 < i < Tencode operator(s) applied in *i*-th step of the plan

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Formula Overview

Formulas with Time Steps

Definition (Time-Stamped Formulas)

Let φ be a propositional logic formula over the variables V. Let 0 < i < T.

We write φ^i for the formula obtained from φ by replacing each $v \in V$ with v^i .

Example: $((a \land b) \lor \neg c)^3 = (a^3 \land b^3) \lor \neg c^3$

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Formula Overview

SAT Formula: Motivation

We want to express a formula whose models are exactly the plans/traces with T steps.

For this, the formula must express four things:

- ▶ The variables v^0 ($v \in V$) define the initial state.
- ▶ The variables v^T ($v \in V$) define a goal state.
- We select exactly one operator variable o^i ($o \in O$) for each time step 1 < i < T.
- ▶ If we select o^i , then variables v^{i-1} and v^i ($v \in V$) describe a state transition from the (i-1)-th state of the plan to the i-th state of the plan (that uses operator o).

The final formula is the conjunction of all these parts.

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Initial State, Goal, Operator Selection

C4.3 Initial State, Goal, Operator Selection

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Initial State, Goal, Operator Selection

SAT Formula: Initial State

SAT Formula: Initial State initial state clauses:

- for all $v \in V$ with $I(v) = \mathbf{T}$
- $ightharpoonup \neg v^0$ for all $v \in V$ with $I(v) = \mathbf{F}$

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Initial State, Goal, Operator Selection

SAT Formula: Goal

SAT Formula: Goal goal clauses:

 $\triangleright \gamma^T$

For STRIPS, this is a conjunction of unit clauses. For general goals, this may not be in clause form.

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Initial State, Goal, Operator Selection

SAT Formula: Operator Selection

Let $O = \{o_1, \dots, o_n\}.$

SAT Formula: Operator Selection operator selection clauses:

operator exclusion clauses:

▶ $\neg o_i^i \lor \neg o_k^i$ for all $1 \le i \le T$, $1 \le j < k \le n$

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C4.4 Transitions

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Transitions

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SAT Formula: Transitions

We now get to the interesting/challenging bit: encoding the transitions.

Key observations: if we apply operator o at time i,

- \triangleright its precondition must be satisfied at time i-1: $o^i \rightarrow pre(o)^{i-1}$
- ▶ variable v is true at time i iff its regression is true at i-1: $o^i \rightarrow (v^i \leftrightarrow regr(v, eff(o))^{i-1})$

Question: Why regr(v, eff(o)), not regr(v, o)?

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Simplifications and Abbreviations

- Let us pick the last formula apart to understand it better (and also get a CNF representation along the way).
- Let us call the formula τ ("transition"): $\tau = o^i \rightarrow (v^i \leftrightarrow regr(v, eff(o))^{i-1}).$
- First, some abbreviations:
 - ightharpoonup Let e = eff(o).
 - Let $\rho = regr(v, e)$ ("regression"). We have $\rho = effcond(v, e) \lor (v \land \neg effcond(\neg v, e))$.
 - Let $\alpha = effcond(v, e)$ ("added").
 - Let $\delta = effcond(\neg v, e)$ ("deleted").

$$\rightarrow \tau = o^i \rightarrow (v^i \leftrightarrow \rho^{i-1}) \text{ with } \rho = \alpha \lor (v \land \neg \delta)$$

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Picking it Apart (1)

Reminder: $\tau = o^i \rightarrow (v^i \leftrightarrow \rho^{i-1})$ with $\rho = \alpha \lor (v \land \neg \delta)$

$$\tau = o^{i} \to (v^{i} \leftrightarrow \rho^{i-1})$$

$$\equiv o^{i} \to ((v^{i} \to \rho^{i-1}) \land (\rho^{i-1} \to v^{i}))$$

$$\equiv \underbrace{(o^{i} \to (v^{i} \to \rho^{i-1}))}_{T_{1}} \land \underbrace{(o^{i} \to (\rho^{i-1} \to v^{i}))}_{T_{2}}$$

 \rightsquigarrow consider this two separate constraints τ_1 and τ_2

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Picking it Apart (2)

Reminder: $\tau_1 = o^i \to (v^i \to \rho^{i-1})$ with $\rho = \alpha \lor (v \land \neg \delta)$

$$\tau_{1} = o^{i} \to (v^{i} \to \rho^{i-1})
\equiv o^{i} \to (\neg \rho^{i-1} \to \neg v^{i})
\equiv (o^{i} \land \neg \rho^{i-1}) \to \neg v^{i}
\equiv (o^{i} \land \neg (\alpha^{i-1} \lor (v^{i-1} \land \neg \delta^{i-1}))) \to \neg v^{i}
\equiv (o^{i} \land (\neg \alpha^{i-1} \land (\neg v^{i-1} \lor \delta^{i-1}))) \to \neg v^{i}
\equiv \underbrace{((o^{i} \land \neg \alpha^{i-1} \land \neg v^{i-1}) \to \neg v^{i})}_{\tau_{11}} \land \underbrace{((o^{i} \land \neg \alpha^{i-1} \land \delta^{i-1}) \to \neg v^{i})}_{\tau_{12}}$$

 \rightsquigarrow consider this two separate constraints τ_{11} and τ_{12}

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Interpreting the Constraints (1)

Can we give an intuitive description of τ_{11} and τ_{12} ? → Yes!

 $\mathbf{r}_{11} = (o^i \wedge \neg o^{i-1} \wedge \neg v^{i-1}) \rightarrow \neg v^i$

"When applying o, if v is false and o does not add it, it remains false."

- called negative frame clause
- ightharpoonup in clause form: $\neg o^i \lor \alpha^{i-1} \lor v^{i-1} \lor \neg v^i$
- $\tau_{12} = (\rho^i \wedge \neg \alpha^{i-1} \wedge \delta^{i-1}) \rightarrow \neg v^i$

"When applying o, if o deletes v and does not add it, it is false afterwards." (Note the add-after-delete semantics.)

- called negative effect clause
- ightharpoonup in clause form: $\neg o^i \lor \alpha^{i-1} \lor \neg \delta^{i-1} \lor \neg v^i$

For STRIPS tasks, these are indeed clauses. (And in general?)

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Picking it Apart (3)

Almost done!

Reminder: $\tau_2 = o^i \rightarrow (\rho^{i-1} \rightarrow v^i)$ with $\rho = \alpha \vee (v \wedge \neg \delta)$

$$\tau_{2} = o^{i} \to (\rho^{i-1} \to v^{i})
\equiv (o^{i} \wedge \rho^{i-1}) \to v^{i}
\equiv (o^{i} \wedge (\alpha^{i-1} \vee (v^{i-1} \wedge \neg \delta^{i-1}))) \to v^{i}
\equiv \underbrace{((o^{i} \wedge \alpha^{i-1}) \to v^{i})}_{\tau_{21}} \wedge \underbrace{((o^{i} \wedge v^{i-1} \wedge \neg \delta^{i-1}) \to v^{i})}_{\tau_{22}}$$

 \rightsquigarrow consider this two separate constraints τ_{21} and τ_{22}

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Interpreting the Constraints (2)

How about an intuitive description of τ_{21} and τ_{22} ?

 $ightharpoonup au_{21} = (o^i \wedge \alpha^{i-1}) \rightarrow v^i$

"When applying o, if o adds v, it is true afterwards."

called positive effect clause

ightharpoonup in clause form: $\neg o^i \lor \neg \alpha^{i-1} \lor v^i$

"When applying o, if v is true and o does not delete it, it remains true."

- called positive frame clause
- ightharpoonup in clause form: $\neg o^i \lor \neg v^{i-1} \lor \delta^{i-1} \lor v^i$

For STRIPS tasks, these are indeed clauses. (But not in general.)

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Summar

C4.5 Summary

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Transitions

SAT Formula: Transitions

SAT Formula: Transitions precondition clauses:

▶ $\neg o^i \lor pre(o)^{i-1}$ for all $1 \le i \le T$, $o \in O$

positive and negative effect clauses:

 $ightharpoonup \neg o^i \lor \neg \alpha^{i-1} \lor v^i$ for all 1 < i < T, $o \in O$, $v \in V$

positive and negative frame clauses:

 $ightharpoonup \neg o^i \lor \neg v^{i-1} \lor \delta^{i-1} \lor v^i$ for all $1 \le i \le T$, $o \in O$, $v \in V$

where $\alpha = effcond(v, eff(o)), \delta = effcond(\neg v, eff(o)).$

For STRIPS, all except the precondition clauses are in clause form. The precondition clauses are easily convertible to CNF

(one clause $\neg o^i \lor v^{i-1}$ for each precondition atom v of o).

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Summai

Summary

- ► SAT planning (planning as satisfiability) expresses a sequence of bounded-horizon planning tasks as SAT formulas.
- Plans can be extracted from satisfying assignments; unsolvable tasks are challenging for the algorithm.
- ► For each time step, there are propositions encoding which state variables are true and which operators are applied.
- We describe a basic sequential encoding where one operator is applied at every time step.
- ▶ The encoding produces a CNF formula for STRIPS tasks.
- ► The encoding follows naturally (with some work) from using regression to link state variables in adjacent time steps.

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