Planning and Optimization C5. SAT Planning: Core Idea and Sequential Encoding

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Planning and Optimization

October 13, 2025 — C5. SAT Planning: Core Idea and Sequential Encoding

C5.1 Introduction

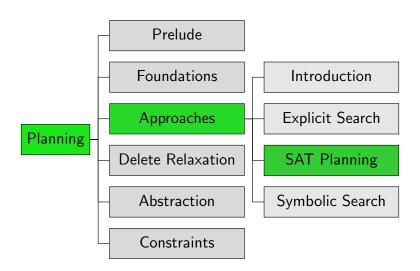
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Content of the Course



C5. SAT Planning: Core Idea and Sequential Encoding

C5.1 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

Complexity Mismatch

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

Solution: Iterative Deepening

- ▶ We can generate a propositional formula that tests if task Π has a plan with horizon (length bound) T in time $O(\|\Pi\|^k \cdot T)$ (\leadsto 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.

SAT Planning: Main Loop

basic SAT planning algorithm:

```
SAT Planning  \begin{aligned} & \mathbf{def} \; \mathsf{satplan}(\Pi) \colon \\ & \quad \mathbf{for} \; T \in \{0,1,2,\dots\} \colon \\ & \quad \varphi := \mathsf{build\_sat\_formula}(\Pi,T) \\ & \quad I = \mathsf{sat\_solver}(\varphi) \qquad \qquad \rhd \; \mathsf{returns} \; \mathsf{a} \; \mathsf{model} \; \mathsf{or} \; \mathsf{none} \\ & \quad \mathsf{if} \; I \; \mathsf{is} \; \mathsf{not} \; \mathsf{none} \colon \\ & \quad \mathsf{return} \; \mathsf{extract\_plan}(\Pi,T,I) \end{aligned}
```

Termination criterion for unsolvable tasks?

C5.2 Formula Overview

SAT Formula: CNF?

- ► 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.

SAT Formula: Variables

- 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 $o \in O$, $1 \le i \le T$ encode operator(s) applied in *i*-th step of the plan

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

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 $1 \le i \le T$.
- 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.

C5.3 Initial State, Goal, Operator Selection

SAT Formula: Initial State

SAT Formula: Initial State initial state clauses:

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

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

SAT Formula: Goal

SAT Formula: Goal goal clauses:



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

SAT Formula: Operator Selection

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

SAT Formula: Operator Selection operator selection clauses:

- operator exclusion clauses:

C5.4 Transitions

SAT Formula: Transitions

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

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

- 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: $o^i \rightarrow (v^i \leftrightarrow \textit{regr}(v, \textit{eff}(o))^{i-1})$

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

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"): $\tau = o^i \to (v^i \leftrightarrow regr(v, eff(o))^{i-1}).$
- First, some abbreviations:
 - $\blacktriangleright \text{ Let } e = eff(o).$
 - 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").
- $\rightarrow \tau = o^i \rightarrow (v^i \leftrightarrow \rho^{i-1}) \text{ with } \rho = \alpha \lor (v \land \neg \delta)$

Picking it Apart (1)

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

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

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

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

 \rightsquigarrow consider this two separate constraints τ_1 and τ_2

Picking it Apart (2)

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

$$\tau_{1} = o^{i} \rightarrow (v^{i} \rightarrow \rho^{i-1})$$

$$\equiv o^{i} \rightarrow (\neg \rho^{i-1} \rightarrow \neg v^{i})$$

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

$$\equiv (o^{i} \land \neg (\alpha^{i-1} \lor (v^{i-1} \land \neg \delta^{i-1}))) \rightarrow \neg v^{i}$$

$$\equiv (o^{i} \land (\neg \alpha^{i-1} \land (\neg v^{i-1} \lor \delta^{i-1}))) \rightarrow \neg v^{i}$$

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

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

Interpreting the Constraints (1)

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

- - "When applying o, if v is false and o does not add it, it remains false."
 - called negative frame clause
 - ▶ in clause form: $\neg o^i \lor \alpha^{i-1} \lor v^{i-1} \lor \neg v^i$

"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
- 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?)

Picking it Apart (3)

Almost done!

Reminder:
$$\tau_2 = o^i \to (\rho^{i-1} \to v^i)$$
 with $\rho = \alpha \lor (v \land \neg \delta)$

$$\tau_2 = o^i \to (\rho^{i-1} \to v^i)$$

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

$$\equiv (o^i \land (\alpha^{i-1} \lor (v^{i-1} \land \neg \delta^{i-1}))) \to v^i$$

$$\equiv \underbrace{((o^i \land \alpha^{i-1}) \to v^i)}_{\tau_{21}} \land \underbrace{((o^i \land v^{i-1} \land \neg \delta^{i-1}) \to v^i)}_{\tau_{22}}$$

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

Interpreting the Constraints (2)

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

- - "When applying o, if o adds v, it is true afterwards."
 - called positive effect clause
 - ▶ 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
 - ▶ 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.)

SAT Formula: Transitions

SAT Formula: Transitions precondition clauses:

- positive and negative effect clauses:
 - ▶ $\neg o^i \lor \neg \alpha^{i-1} \lor v^i$ for all $1 \le i \le T$, $o \in O$, $v \in V$
- positive and negative frame clauses:

 - $ightharpoonup \neg o^i \lor \alpha^{i-1} \lor v^{i-1} \lor \neg v^i$ 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).

C5. SAT Planning: Core Idea and Sequential Encoding

Summary

C5.5 Summary

Summary: Sequential SAT Encoding (1)

Sequential SAT Encoding (1)

initial state clauses:

$$\triangleright$$
 v^0

for all
$$v \in V$$
 with $I(v) = \mathbf{T}$

$$\rightarrow \neg v^0$$

for all
$$v \in V$$
 with $I(v) = \mathbf{F}$

goal clauses:

$$ightharpoonup \gamma^T$$

operator selection clauses:

$$\triangleright$$
 $o_1^i \vee \cdots \vee o_n^i$

for all
$$1 \le i \le T$$

operator exclusion clauses:

$$ightharpoonup \neg o_i^i \lor \neg o_k^i$$

for all
$$1 \le i \le T$$
, $1 \le j < k \le n$

Summary: Sequential SAT Encoding (2)

Sequential SAT Encoding (2)

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:

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

positive and negative frame clauses:

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

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

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

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.