

Artificial Intelligence

Classical Planning: Planning via Inference

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Recap: Classical Planning

Classical planning seeks a path from the initial state to a goal through a **finite, deterministic, fully-observable** search space. We describe a classical planning task using PDDL (STRIPS) syntax:

PDDL domain description:

- States (predicates)
- Action scheme a
 - Parameters
 - Preconditions $Precond(a)$
 - Effects $Effect(a)$: $Add(a)$, $Del(a)$

PDDL problem description:

- Initial state I
- Goal g

Our aim. Building a classic planner

Challenge. Understand the task's relations with **search** and **logic**.

Main paradigms:

- Search-based (covered in the last lecture)
 - Forward (Progression) planning
 - Backward (Regression) planning
- Logic-based (to be covered in this lecture)
 - Propositional logic-based planning: SATPlan
 - Logic programming-based planning: Prolog planner

Logic-based Planning

Planning v.s. Logic:

- **States** can be expressed as logical sentences.
- **Actions** can be expressed as logical rules that describe the effects thereby capturing state transitions.
- **Goal** is true only if a sequence of actions are true, triggering a sequence of state transitions that link the initial state with the goal.
- **Planning task** can be expressed as a **knowledge base**:

$$\Phi = \text{initial state} \wedge \text{action descriptions} \wedge \text{goal}$$

- **Planning** is equivalent to checking **satisfiability** of Φ , i.e., finding an interpretation π such that

$$\pi \models \Phi$$

	Logic	Planning
States	Logic sentences	Logic sentences
Actions	Logic rules	Preconditions/effects
Goals	Logical sentences	Logical sentence
Plan	Satisfying interpretation	Sequence of actions

Propositional Logic-based Planner: SATPlan

Recall: **Propositional satisfaction problem (SAT)**

INPUT A set of propositions Φ

OUTPUT Find an interpretation (that determines the truth values of atomic propositions) π such that $\pi \models \Phi$; *failure* if such an π does not exist.

SAT Solvers^a: DPLL algorithm, local search, etc.

E.g. $(p \vee \neg q) \wedge (\neg p \vee q)$ is satisfiable by $\pi(p) = 1$ and $\pi(q) = 1$.

$(p \vee \neg q) \wedge (\neg p \vee q) \wedge (\neg p \vee \neg q) \wedge (p \vee q)$ is not satisfiable.

^aBeyond the scope of this course.

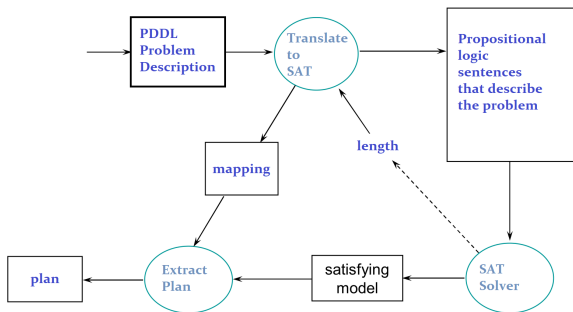
SATPlan¹ finds a plan by converting the problem to a propositional KB Φ :

- A satisfying interpretation of Φ : assign *true* to the actions that are part of a correct plan; *false* to the others.
- If there is no correct plan, Φ is not satisfiable.

¹We mentioned it in our propositional logic lectures.

Main components of SATPlan:

- 1 **TranslateToSAT**: Translate a PDDL description into a proposition KB.
- 2 **SATSolver**: Feed this propositions to a SAT solver.
 - If the sentence is unsatisfiable, then there is not valid plan.
 - If a satisfying interpretation is found, then the goal can be achieved.
- 3 **ExtractPlan**: If the goal can be achieved, extract action variables at each time $1 \leq i \leq t$ to form a plan.



- Index time steps by $t = 0, 1, 2, 3, \dots$
- A proposition needs to have a finite length.
- Set a hyperparameter T_{\max} to bound the length of the plan.

A **bounded planning problem** is a pair $(problem, t)$ where

- *problem* is a planning problem; t is a positive integer.
- A solution is a correct plan for *problem* that has length t .

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SATPlan procedures: Iterative deepening

for $t = 0, 1, 2, \dots, T_{\max}$

- **TranslateToSAT:** Encode $(problem, t)$ as a set of propositions Φ
 - **SATSolver:** Solve SAT on Φ
 - **if** Φ is satisfiable **then**
 - **ExtractPlan:** Construct a plan from the satisfying interpretation
 - **end if**
- end for**

Question. How to translate a PDDL description to SAT?

We illustrate the process using a concrete example.

Example. [spare tire]

Action(Remove(obj), PRECOND: AtAxle(obj),

EFFECT: \neg AtAxle(obj) \wedge AtGround(obj))

Action(PutOn(obj), PRECOND: AtGround(obj) \wedge \neg AtAxle(Flat),

EFFECT: \neg AtGround(obj) \wedge AtAxle(obj))

Init(AtAxle(Flat) \wedge AtGround(Spare))

Goal(AtAxle(Spare))



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Goal(AtAxle(Spare))

Main steps of **TranslateToSAT**:

- ① **Preparation:** propositionalise actions.
- ② **Code initial state**
- ③ **Code goal**
- ④ **Precondition axioms**
- ⑤ **Successor-state axioms**
- ⑥ **Action exclusion axioms**



- **Step i). Propositionalise actions:** Instantiate each action:

Action(Remove(Flat),PRECOND: AtAxle(Flat),

EFFECT: \neg AtAxle(Flat) \wedge AtGround(Flat))

Action(Remove(Spare),PRECOND: AtAxle(Spare),

EFFECT: \neg AtAxle(Spare) \wedge AtGround(Spare))

Action(PutOn(Flat),PRECOND: AtGround(Flat) \wedge \neg AtAxle(Flat),

EFFECT: \neg AtGround(Flat) \wedge AtAxle(Flat))

Action(PutOn(Spare),PRECOND: AtGround(Spare) \wedge \neg AtAxle(Flat),

EFFECT: \neg AtGround(Spare) \wedge AtAxle(Spare))

Create a new proposition for every **ground term**, **ground action**, and time step $i = 0, \dots, t$:

Remove_Flatⁱ, AtAxle_Flatⁱ, AtGround_Flatⁱ, Remove_Spareⁱ,

AtAxle_Spareⁱ, AtGround_Spareⁱ, PutOn_Flatⁱ, \dots , Puton_Spareⁱ

The translation consists of the following:

- **Step ii). Initial state:** Assert propositions for every positive literal appear/not appear in the initial state.

$$AtAxle_Flat^0 \wedge AtGround_Spare^0 \wedge \neg AtGround_Flat^0 \wedge \neg AtAxle_Spare^0$$

- **Step iii). Goal:** For every variable in the goal, replace the literal that contain the variable with a disjunction over constants.

$$AtAxle_Spare^t$$

- **Step iv). Precondition axioms:** For each ground action A , add the axiom $A^{i+1} \rightarrow Precond(A)^i$ for $i = 0, 1, \dots, t - 1$.

$$Remove_Flat^{i+1} \rightarrow AtAxle_Flat^i$$

$$Remove_Spare^{i+1} \rightarrow AtAxle_Spare^i$$

$$PutOn_Flat^{i+1} \rightarrow (AtGround_Flat^i \wedge \neg AtAxle_Flat^i)$$

$$PutOn_Spare^{i+1} \rightarrow (AtGround_Spare^i \wedge \neg AtAxle_Flat^i)$$

- **Step v). Successor-state axioms:** For each positive literal F , add $i = 1, \dots, t$

$$F^i \leftrightarrow \text{ActionCauses}F^i \vee (F^{i-1} \wedge \neg \text{ActionCausesNot}F^i),$$

where $\text{ActionCauses}F$ is a disjunction of all the ground actions that have F in their add list, and $\text{ActionCausesNot}F$ is a disjunction of all the ground actions that have F in their delete list.

$$\text{AtAxle_Flat}^i \leftrightarrow (\text{PutOn_Flat}^i \vee (\text{AtAxle_Flat}^{i-1} \wedge \neg \text{Remove_Flat}^i)),$$

$$\text{AtGround_Flat}^i \leftrightarrow (\text{Remove_Flat}^i \vee (\text{AtGround_Flat}^{i-1} \wedge \neg \text{PutOn_Flat}^i)),$$

$$\text{AtAxle_Spare}^i \leftrightarrow (\text{PutOn_Spare}^i \vee (\text{AtAxle_Spare}^{i-1} \wedge \neg \text{Remove_Spare}^i)),$$

$$\text{AtGround_Spare}^i \leftrightarrow (\text{Remove_Spare}^i \vee (\text{AtGround_Spare}^{i-1} \wedge \neg \text{PutOn_Spare}^i)),$$

- **Step vi). Action exclusion axioms:** Actions are not taken at the same time.

For any $i = 1, \dots, t$

$$\neg \text{Remove_Flat}^i \vee \neg \text{Remove_Spare}^i$$

$$\neg \text{Remove_Flat}^i \vee \neg \text{PutOn_Flat}^i$$

$$\neg \text{Remove_Flat}^i \vee \neg \text{PutOn_Spare}^i$$

$$\neg \text{Remove_Spare}^i \vee \neg \text{PutOn_Flat}^i$$

$$\neg \text{Remove_Spare}^i \vee \neg \text{PutOn_Spare}^i$$

$$\neg \text{PutOn_Flat}^i \vee \neg \text{PutOn_Spare}^i$$

A satisfying interpretation:

The following atoms are true (the rest are false).

- Time step 0 (initial state):

$AtAxle_Flat^0, AtGround_Spare^0,$

- Time step 1:

$Remove_Flat^1, AtGround_Flat^1, AtGround_Spare^1,$

- Time step 2:

$PutOn_Spare^2, AtAxle_Spare^2, AtGround_Flat^2.$

Thus a plan is found with length $t = 2$:

1.*Remove(Flat),* 2.*PutOn(Spare).*

Online demo: We can manually input a SAT and construct a plan.

- **Online DPLL SAT solver:**

<https://www.inf.ufpr.br/dpasqualin/d3-dpll/>

Input a set of propositions in **conjunctive normal form**.

Solve SAT and find a satisfying interpretation.

- **Python Boolean library (PBL) which implements a CNF converter:** <https://github.com/tyler-utah/PBL>

DPLL SAT Solver

This version of DPLL implements unit clause and non-chronological backtrack. The assignment is in lexicographical order.

Enter in the box below a series of clauses (one for each line), using alphanumeric characters to represent the variables, separating it using spaces. A dash (-) represents the negation symbol.

```
AA0  
AG0  
-AG0  
-AA0  
AA0  
-RM1  
-RM1 AA0  
-RM1 AA0  
-POF1 AG0  
-POF1 -AA0  
-POF1 AG0  
-POF1 -AA0
```

☐ Step by Step?

Solution: SATISFIABLE AA0 AG0 AA0 RM1 AG1 POS2 AG1 AG2 -AG0 -AA0 -RM1
-POF1 -POF1 -AA0 -RM1 -RM1 -AA0 -POF1 -AA0



This is the online interface for PBL, a Boolean algebra/propositional logic library written in Python.

[About the PBL package](#)

[Language Spec](#)

ENTER YOUR FORMULA HERE

#Example showing some of the propagation simplifications.
why propagating 0 into variable x4, the whole formula should
at one out to be false.

#NOTE: the simplifications are not very robust and just
#approximate when they are obvious.

Main_Exp (AG02 <=> (RM02 | (AG01 & -PO02)))

PROGRAM OUTPUT

(-AG02 | RM02 | AG01) & (-AG02 | RM02 | -PO02) & (-RM02 |
AG02) & (-AG01 | PO02 | AG02)

Disadvantage: A planning problem usually requires a large propositional KB.

TranslateToSAT needs to create

- $(T_{\max} + 1) \times |Obj|^{Arg_{sp}}$ new atomic propositions for each predicate symbol, and
- $T_{\max} \times |Obj|^{Arg_{sA}}$ new atomic propositions for each action schema,

where $|Obj|$ is the set of constants, Arg_{sp} is the maximum arity of a predicate, Arg_{sA} is the maximum arity of an action scheme.

Advantages: Speed

- Utilising efficient domain-independent heuristic for propositional logic reasoning
- Taking advantage of mature SAT solver such as DPLL, which is highly optimised.
- Fixed structure in classical planning domain and problem means that further optimisation is possible.

Logic Programming-base Planner

Idea: Write a planner in **Prolog**

- PDDL and logic programs are both **descriptive** languages.
- Instead of using propositional logic, we write a **logic program** that defines valid plans.
- **Challenge:** Translating PDDL to Prolog.
- **Assumption:** Initial state, goal, and preconditions do not contain negative literals.

Advantage:

- Taking advantage of Prolog interpreter which is highly optimised.
- We are essentially developing a “PDDL interpreter”:
 - Define **domain-independent** Prolog rules to for planning
 - Describe PDDL domain using Prolog facts.

We need to define how Prolog can code the following **main components (Domain independent)**:

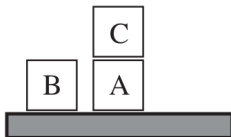
- ① Predicates
- ② States
- ③ Actions
- ④ Plan
- ⑤ Goal
- ⑥ **Change State**: State transitions (current state, plan, next state).
- ⑦ **Conditions met**: Whether a state satisfies the precondition of an action.
- ⑧ **Goal met**: Whether a state satisfies the goal.

We next present a Prolog planner that is developed by Luger, Stubblefield, and Davis at UNM at <https://www.cs.unm.edu/~luger/ai-final/code/PROLOG.planner.html>. Other Prolog planner is also possible².

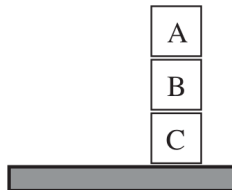
²See *Prolog: Programming for artificial intelligence*, by I. Bratko.

Example. [Blocks world domain]

- Predicates: $onTable(x)$, $on(x, y)$, $clear(x)$
- Action $moveToTable(x, y)$
 - Preconditions: $clear(x) \wedge on(x, y)$
 - Effects: $clear(y) \wedge onTable(x) \wedge \neg on(x, y)$
- Action $moveToBlock1(x, y, z)$
 - Preconditions: $clear(x) \wedge clear(z) \wedge on(x, y)$
 - Effects: $clear(y) \wedge on(x, z) \wedge \neg clear(z) \wedge \neg on(x, y)$
- Action $moveToBlock2(x, y)$
 - Preconditions: $clear(x) \wedge clear(y) \wedge onTable(x)$
 - Effects: $on(x, y) \wedge \neg clear(y) \wedge \neg onTable(x)$



Start State



Goal State

- **1. Predicate:** A predicate is represented by a function symbol.

- terms represent literals.
- ground terms represent ground literals.

E.g. `on(X,Y)`, `clear(X)`, `ontable(a)`, `clear(a)`

- **2. States:** A state is a **set** of ground terms.

E.g. `[ontable(b), on(c,a), ontable(a), clear(b), clear(a)]`

- **3. Actions:** An action name is represented by a function symbol.

- Precondition: a set of terms
- Effects: a set of Del and Add terms.

E.g.

```
act(movetotable(X, Y), [clear(X), on(X,Y)],  
    [del(on(X,Y)), add(clear(Y)), add(ontable(X))]).
```

- **4. Plan:** A plan is represented by a list of ground actions.

E.g. `[movetotable(c,a), movetoblock2(b,c), movetoblock1(a,b)]`

- **5. Goal:** A goal is represented by a **set** of ground terms.

E.g. `[ontable(c), on(a,b), on(b,c), clear(a)]`

● **6. Change state** Parameters: S (current state), $Effects$, S' (new state)

- **Case 1.** If $Effects$ is empty, then new state $S' = S$.

`change_state(S, [], S).`

- **Case 2.** If the first effect is in *Add*, then add it to the state.

`change_state(S, [add(P) | T], S_new) :- change_state(S, T, S2),
add_to_set(P, S2, S_new), !.`

- **Case 3.** If the first effect is in *Del*, then remove it from the state.

`change_state(S, [del(P) | T], S_new) :- change_state(S, T, S2),
remove_from_set(P, S2, S_new), !.`

● **7. Condition met**

`conditions_met(Precond, State) :- subset(Precond, State).`

● **8. Goal met**

`goal_met(State, Goal) :- equal_set(State, Goal).`

Note. Here we use a Prolog implementation of *set data structure* with operations such as `subset`, `add_to_set`, `remove_from_set` and `equal_set`.

Progression planning on Prolog:

```
plan(State, Goal, _, Plan) :- goal_met(State, Goal),  
    write('actions are'), nl,  
    reverse_print_stack(Plan).
```

```
plan(State, Goal, Been_list, Plan) :-  
    act(Name, Preconditions, Effects),  
    conditions_met(Preconditions, State),  
    change_state(State, Effects, Child_state),  
    not(member_state(Child_state, Been_list)),  
    stack(Child_state, Been_list, New_been_list),  
    stack(Name, Plan, New_plan),  
    plan(Child_state, Goal, New_been_list, New_plan),!.
```

```
go(S, G) :- plan(S, G, [S], []).
```

Prolog domain specification:

%domain definition

```
act(movetoblock2(X, Y), [clear(X), clear(Y), ontable(X)],  
    [del(ontable(X)), del(clear(Y)), add(on(X,Y))]).
```

```
act(movetoblock1(X, Y, Z), [clear(X), clear(Z), on(X,Y)],  
    [del(clear(Z)), del(on(X,Y)), add(clear(Y)), add(on(X,Z))]).
```

```
act(movetotable(X, Y), [clear(X), on(X,Y)],  
    [del(on(X,Y)), add(clear(Y)), add(ontable(X))]).
```

%problem definition

%initial and goal states

```
test :- go([ontable(a), ontable(b), clear(b), clear(c), on(c,a)],  
    [clear(a),on(a,b),on(b,c),ontable(c)]).
```

A run-through:

① Start state: [ontable(a), ontable(b), clear(b), clear(c), on(c,a)]

Goal: [clear(a), on(a,b), on(b,c), ontable(c)]

② movetoblock1(c,a,b):

State: [ontable(a), ontable(b), clear(c), clear(a), on(c,b)]

③ movetotable(c,b):

State: [ontable(a), ontable(b), clear(c), clear(a), clear(b), ontable(c)]

④ movetoblock2(b,a):

State: [ontable(a), clear(c), clear(b), ontable(c), on(b,a)]

⑤ movetoblock1(b,a,c):

State: [ontable(a), clear(b), ontable(c), clear(a), on(b,c)]

⑥ movetoblock2(a,b):

State: [ontable(c), clear(a), on(b,c), on(a,b)]

Note. This is a progression planner.

Summary of The Topic

The following are the main knowledge points covered:

- **Logic-based Planning:**
 - Making use of well-developed and highly-optimised logic inference mechanisms.
 - Reducing planning to the satisfaction problem.
- **Propositional logic-based planner: SATPlan**
 - Main components: TranslateToSAT, SATSolver, ExtractPlan
 - Translate PDDL to propositional KB
- **Logic programming-based planner: Prolog planner**
 - States and goal expressed as sets.
 - Progression planner implementation.