Planning

Chapter 10

Outline

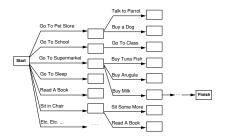
- Search vs. planning
- Using PDDL for planning

Search vs. planning

• Consider the task get milk, bananas, and a cordless drill

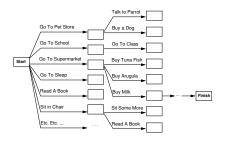
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- Problems:
 - Enormous search space
 - Actions are complex objects:
 - They have preconditions and they change the world
 - Simple goal test is inadequate



Search vs. planning contd.

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- 2 divide-and-conquer by subgoaling

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Compare:

	Search	Planning
States	Data structures	Logical sentences
		(positive ground literals)
Actions	Code	Preconditions/outcomes
		in a schema
Goal	Test	Logical sentence
		(conjunction of literals)
Plan	Sequence from S_0	Sequence from S_0

PDDL (= Planning Domain Definition Language) is a planning system derived from STRIPS, which goes back to 1971(!).

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 - An action instance transforms the world description.
- Given an *initial world description*, find a sequence of action instances that achieves a given *goal*.
- No explicit mention is made of time.



World States

- The world or domain is described as a variable-free set of atomic formulas.
- Example:

```
\{Block(a), Block(b), \ldots, On(a, b), OnTable(b), \ldots, Clear(c), \ldots\}
```

- Uses database semantics: If a fact doesn't appear in the list, it is assumed to be false.
 - E.g. If On(b,c) isn't in the domain description $\neg On(b,c)$ is assumed to hold.
- Constants are assumed to denote distinct individuals, i.e. $a \neq b$.

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- E.g.: Action(Fly(p, from, to)
 PRECOND:
 At(p, from) ∧ Flight(p) ∧ Airport(from) ∧ Airport(to)
 EFFECT: ¬At(p, from) ∧ At(p, to))

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- Stack(x, y): Move x from being on the table to being on y.
 - PRECOND: $OnTable(x) \land Clear(x) \land Clear(y) \land x \neq y$
 - EFFECT: $On(x, y) \land \neg OnTable(x) \land \neg Clear(y)$

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- *Unstack*: (Exercise)

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• To establish a goal, a sequence of action instances needs to be found that leads from the initial state to the goal.

- An action instance a is an action along with bindings for its free variables.
- E.g. recall the schema:
 Action(Fly(p, from, to)
 PRECOND:
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    E.g. recall the schema:

  Action(Fly(p, from, to)
    PRFCOND:
       At(p, from) \land Flight(p) \land Airport(from) \land Airport(to)
    EFFECT: \neg At(p, from) \land At(p, to)
  This has instance:
  Action(Fly(AC118, YVR, YYZ)
    PRECOND:
       At(AC118, YVR) \land Flight(AC118) \land
            Airport(YVR) \land Airport(YYZ)
    EFFECT: \neg At(AC118, YVR) \land At(AC118, YYZ)
```

- An action instance a is possible in state s iff every precondition in PRECOND(a) holds in s.
- If we describe s by listing those atoms that hold in s, then this can be expressed as
 - $PRECOND^+(a) \subseteq s$
 - $PRECOND^-(a) \cap s = \emptyset$

where

- PRECOND⁺(a) is the set of positive literals and
- $PRECOND^{-}(a)$ is the set of negated literals

in the precondition.

Equivalently, we can write:

$$PRECOND(a) \subseteq s \cup \{ \neg p \mid p \notin s \}.$$

- Let
 - ADD(a) be the set of positive literals in EFFECT(a) and
 - DEL(a) be the set of atoms given by the negative literals in EFFECT(a).
- The result of executing an action instance a that is possible in s is the state:

$$RESULT(a, s) = (s - DEL(a)) \cup ADD(a).$$

• Given an instantiated action sequence a_1, \ldots, a_n , and a situation s, we set

$$S_0 = s$$

and

$$s_i = RESULT(a, s_{i-1})$$
 for $i = 1, ..., n$.

- The action sequence succeeds if every individual action succeeds.
- The action sequence achieves the goal G if s_n entails G.

- Planning can be done in either a "forward" or "backward" manner.
- Known as *progressive* and *regressive* planning respectively.
- Originally regressive planners were most used, due to their focus on the goal
- With better heuristics and increased computational power, progressive planners have come to dominate.

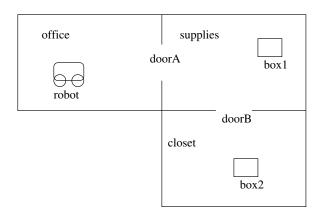
Progressive Planning in PDDL

- The most intuitive way to try to obtain a plan is to:
 - · begin at the initial state and
 - find a sequence of actions that lead to the goal.
- This is called a *progressive planner* since it progresses the initial state forward until a state satisfying the goal is found.

Progressive Planning

Depth-First Progressive Planner: Input: A world description S and goal formula Goal Output: A plan or fail ProgPlan(S, Goal) if Goal \subseteq S then return empty plan for each operator instance (Act, Pre, Add, Del) such that S satisfies Pre do { let $S' = (S \setminus Del) \cup Add$ let Plan = ProgPlan(S', Goal)if Plan \neq fail then return Plan-Act return fail

Example



Goal: Get some box into the office

Example

Initial world DB:

```
Box(box1), Box(box2),
```

InRoom(box1, supplies), InRoom(box2, closet),
InRoom(robot, office),

Connected(office, supplies), Connected(supplies, office), Connected(closet, supplies), Connected(supplies, closet)

Example

Action schema:

```
goThru(r1, r2)
```

- PRECOND: InRoom(robot, r1), Connected(r1, r2)
- EFFECT: InRoom(robot, r2), ¬InRoom(robot, r1),

pushThru(x, r1, r2)

- PRECOND: InRoom(robot, r1), InRoom(x, r1), Connected(r1, r2)
- EFFECT: InRoom(robot, r2), InRoom(x, r2),
 ¬InRoom(robot, r1), ¬InRoom(x, r1)

Progressive Planning Example

```
With goThru(office, supplies), obtain first progressed DB:
    Box(box1), Box(box2),
    InRoom(box1, supplies), InRoom(box2, closet),
         InRoom(robot, supplies),
    Connected (office, supplies), Connected (supplies, office),
     Connected (closet, supplies), Connected (supplies, closet)
With pushThru(box1, supplies, office), obtain the DB:
    Box(box1), Box(box2),
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         InRoom(robot, office),
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Regressive Planning with PDDL

- Idea: Begin with the goal state, and work backwards to try to get to the initial state.
- The search space can be defined in a "backwards chaining" fashion:
- Idea: Work backwards, repeatedly simplifying the goal until we get a goal satisfied in the initial state.
- Called goal regression

Regressive Planning

Depth-First Regressive Planner:

```
Input: The initial world description Init and a goal formula Goal Output: A plan or fail
```

```
RegrPlan(Init, Goal) if Goal \subseteq Init then return empty plan for each operator instance \langle Act, Pre, Add, Del \rangle such that Del \cap Goal = \emptyset { let Goal' = (Goal \cup Pre) \setminus Add let Plan = RegrPlan(Init, Goal') if Plan \neq fail then return Plan \cdot Act } return fail
```

Regressive Planning Example

- Planner is called with the initial world DB and the goal: Box(x), InRoom(x, office)
- The goal is not satisfied by the initial world DB.
- The action instance pushThru(box1, supplies, office)

 has a delete list that does not intersect with the goal.
- Get regressed subgoal:
 Box(box1), InRoom(robot, supplies), InRoom(box1, supplies),
 Connected(supplies, office)
- The action instance: goThru(office, supplies)
 yields the regressed goal:
 Box(box1), InRoom(robot, office), InRoom(box1, supplies),
 Connected(supplies, office), Connected(office, supplies)
- This is satisfied in the initial state.

- A, B, and C are on the table.
- The goal is On(A, B) and On(B, C).
- Initial state:
 Init = {OnTable(A), OnTable(B), OnTable(C),
 Clear(A), Clear(B), Clear(C)}

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- Also: domain-specific heuristics

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Other types of planners:

- Partial-order planners
- GRAPHPLAN

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- General planning comment: Things get tricky very quickly.
 - E.g. On(B, table), On(C, A), Goal: On(A, B), On(B, C).

