

PLANNING AND SEARCH

CLASSICAL PLANNING

Outline

- ◇ Search vs. planning
- ◇ STRIPS operators
- ◇ PDDL
- ◇ Forward (progression) state-space search
- ◇ Backward (regression) relevant-states search
- ◇ Partial-order planning

Search vs. planning contd.

Planning systems do the following:

- 1) open up action and goal representation to allow selection
- 2) divide-and-conquer by subgoaling
- 3) relax requirement for sequential construction of solutions

	Search	Planning
States	data structures	Logical sentences
Actions	code	Preconditions/outcomes
Goal	code	Logical sentence (conjunction)
Plan	Sequence from S_0	Constraints on actions

Classical planning

Assumptions are:

- (1) Environment is deterministic
- (2) Environment is observable
- (3) Environment is static (it only in response to the agent's actions)

STRIPS operators

STRIPS planning language (Fikes and Nilsson, 1971)

Tidily arranged actions descriptions, restricted language

ACTION: $Buy(x)$

PRECONDITION: $At(p), Sells(p, x)$

EFFECT: $Have(x)$

[Note: this abstracts away many important details!]

Restricted language \Rightarrow efficient algorithm

Precondition: conjunction of positive literals

Effect: conjunction of literals

$At(p) \ Sells(p, x)$

Buy(x)

$Have(x)$

PDDL

Planning Domain Definition Language

A bit more relaxed than STRIPS

Preconditions and goals can contain negative literals

ACTION: $Buy(x)$

PRECONDITION: $At(p), Sells(p, x)$

EFFECT: $Have(x)$

is called an **action schema**

Planning domain

States are sets of fluents (ground, functionless atoms). Fluents which are not mentioned are false. (Closed world assumption.)

$a \in \text{ACTIONS}(s)$ iff $s \models \text{PRECOND}(a)$

$\text{RESULT}(s, a) = (s - \text{DEL}(a)) \cup \text{ADD}(a)$

where $\text{DEL}(a)$ is the list of literals which appear negatively in the effect of a , and $\text{ADD}(a)$ is the list of positive literals in the effect of a .

Example (slightly modified)

ACTION: $Buy(x)$

PRECONDITION: $At(p), Sells(p, x), Have(Money)$

EFFECT: $Have(x), \neg Have(Money)$

$DEL(Buy(Jaguar)) = \{Have(Money)\}$

$ADD(Buy(Jaguar)) = \{Have(Jaguar)\}$

If $s = \{At(JDealer), Sells(JDealer, Jaguar), Blue(Sky), Have(Money)\}$,

$Buy(Jaguar) \in ACTIONS(s)$

$RESULT(s, Buy(Jaguar)) = (s - \{Have(Money)\}) \cup \{Have(Jaguar)\}$

$= \{At(JDealer), Sells(JDealer, Jaguar), Blue(Sky), Have(Jaguar)\}$

Planning problem

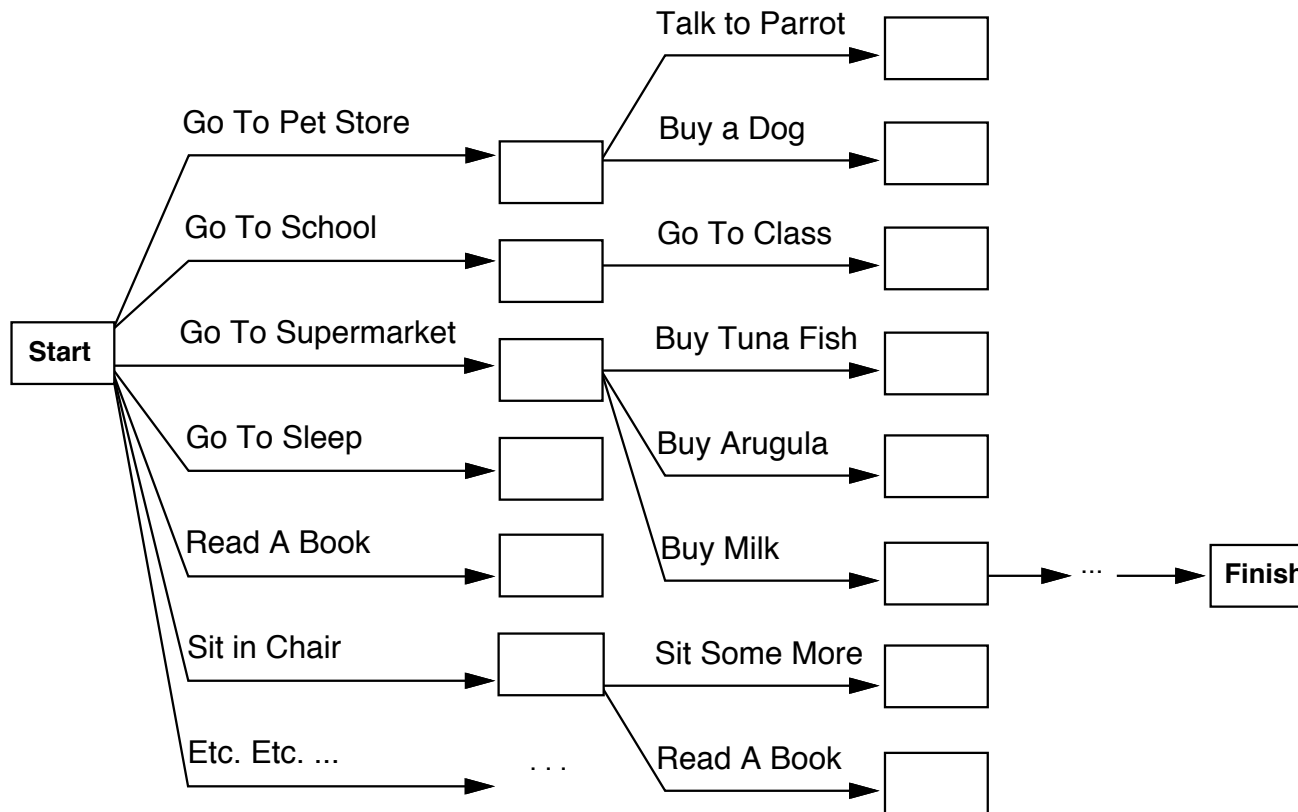
Planning problem = planning domain + initial state + goal

Goal is a conjunction of literals: $Have(Jaguar) \wedge \neg At(Jail)$

Can solve planning problem using search

Forward (progression) planning

Searching for a solution starting from the initial state looks hopeless



Forward planning 2

However, it turns out we can automatically derive good heuristics (and remember how much better A^* is compared to uninformed search)

Two basic approaches:

1) add more edges to the graph (make more actions possible), and use solutions to the resulting problem as a heuristic

Examples: remove (some) preconditions, ignore delete lists...

ACTION $Slide(t, s_1, s_2)$

PRECOND: $On(t, s_1) \wedge Tile(t) \wedge Blank(s_2) \wedge Adjacent(s_1, s_2)$

EFFECT: $On(t, s_2) \wedge Blank(s_1) \wedge \neg On(t, s_1) \wedge \neg Blank(s_2)$

removing $Blank(s_2)$ will enable tiles to move to occupied places: Manhattan distance heuristic

2) abstract the problem (make the search space smaller).

Backward (regression) planning

Also called **relevant-states** search

Start at the goal state(s) and do **regression** (go back):

Given a goal description g and a ground action a , the regression from g over a gives a state description g' :

$$g' = (g - \text{ADD}(a)) \cup \{\text{PRECOND}(a)\}$$

For example, if the goal is $\text{Have}(\text{Jaguar}) \wedge \neg \text{At}(\text{Jail})$,

$$g' = (\{\text{Have}(\text{Jaguar}), \neg \text{At}(\text{Jail})\} - \{\text{Have}(\text{Jaguar})\}) \cup$$

$$\{\text{At}(p), \text{Sells}(p, \text{Jaguar}), \text{Have}(\text{Money})\} =$$

$$\{\neg \text{At}(\text{Jail}), \text{At}(p), \text{Sells}(p, \text{Jaguar}), \text{Have}(\text{Money})\}$$

note that g' is partially uninstantiated (p is a free variable)

Backward (regression) planning 2

Which actions to regress over?

Relevant actions: have an effect which is in the set of goal elements and no effect which negates an element of the goal.

For example, *Buy(Jaguar)* is a relevant action. *Steal(Jaguar)* may also result in *Have(Jaguar)* but if it has an additional effect of *At(Jail)*, it is not a relevant action.

Search backwards from g , remembering the actions and checking whether we reached an expression applicable to the initial state.

A lot fewer actions/relevant states than forward search, but uses sets of states (g, g') - hard to come up with good heuristics.

Totally vs partially ordered plans

So far we produced a linear sequence of actions (totally ordered plan)

Often it does not matter in which order *some of the actions* are executed

For problems with independent subproblems often easier to find a **partially ordered plan**: a plan which is a set of actions and a set of constraints

$Before(a_i, a_j)$

Partially ordered plans are created by a search through a space of plans (rather than the state space)

Next lecture

More classical planning

