Foundations of Artificial Intelligence 10. Action Planning

Solving Logically Specified Problems using a General Problem Solver

Joschka Boedecker and Wolfram Burgard and Frank Hutter and Bernhard Nebel and Michael Tangermann



Albert-Ludwigs-Universität Freiburg

June 7, 2019

Contents

- What is Action Planning?
- Planning Formalisms
- Basic Planning Algorithms
- Computational Complexity
- Current Algorithmic Approaches
- Current Trends in Planning
- Summary

Lecture Overview

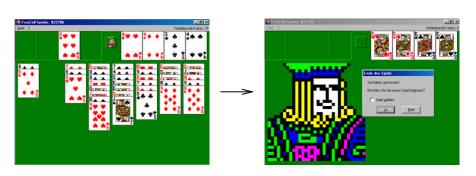
- What is Action Planning?
- Planning Formalisms
- Basic Planning Algorithms
- Computational Complexity
- Current Algorithmic Approaches
- Current Trends in Planning
- Summary

Planning

- Planning is the art and practive of thinking before acting [Haslum]
- Planning is the process of generating (possibly partial) representations of <u>future behavior</u> prior to the use of such plans to constrain or control that behavior.
- The outcome is usually a <u>set of actions</u>, with temporal and other constraints on them, for execution by some agent or agents.

Planning Tasks

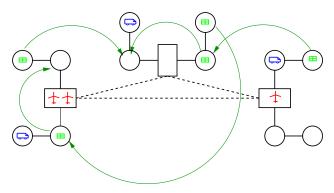
Given a current state, a set of possible actions, a specification of the goal conditions, which plan transforms the current state into a goal state?



Search through possible configuration

Another Planning Task: Logistics

Given a road map, and a number of trucks and airplanes, make a plan to transport objects from their start to their goal destinations.



Action Planning is not ...

- Problem solving by search, where we describe a problem by a state space and then implement a program to search through this space
 - in action planning, we specify the problem declaratively (using logic) and then solve it by a general planning algorithm
- Program synthesis, where we generate programs from specifications or examples
 - in action planning we want to solve just one instance and we have only very simple action composition (i.e., sequencing, perhaps conditional and iteration)
- Scheduling, where all jobs are known in advance and we only have to fix time intervals and machines
 - instead we have to find the right actions and to sequence them
- Of course, there is interaction with these areas!

Lecture Overview

- What is Action Planning?
- Planning Formalisms
- Basic Planning Algorithms
- Computational Complexity
- Current Algorithmic Approaches
- Current Trends in Planning
- Summary

Domain-Independent Action Planning

- Start with a declarative specification of the planning problem
- Use a domain-independent planning system to solve the planning problem
- Domain-independent planners are generic problem solvers
- Issues:
 - Good for evolving systems and those where performance is not critical
 - Running time should be comparable to specialized solvers
 - Solution quality should be acceptable
 - ... at least for all the problems we care about

Planning as Logical Inference

Planning can be elegantly formalized with the help of the *situation* calculus.

Initial state:

 $(At(truck1, loc1, s_0) \land (At)(package1, loc3, s_0))$

Operators (successor-state axioms):

 $\forall a, s, l, p, t \ At(t, p, Do(a, s)) \Leftrightarrow \{a = Drive(t, l, p) \land Poss(Drive(t, l, p), s) \lor At(t, p, s) \land (a \neq \neg Drive(t, p, l, s) \lor \neg Poss(Drive(t, p, l), s))\}$

Goal conditions (query):

 $\exists s \ At(package1, loc2, s)$

The constructive proof of the existential query (computed by a automatic theorem prover) delivers a plan that does what is desired.

Can be quite inefficient!

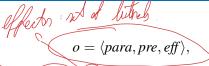
The Basic STRIPS Formalism

STRIPS: STanford Research Institute Problem Solver

- S is a *first-order vocabulary* (predicate and function symbols) and Σ_S denotes the set of *ground atoms* over the signature (also called **facts** or **fluents**).
- $\Sigma_{S,V}$ is the set of atoms over S using variable symbols from the set of variables V.
- A first-order STRIPS state S is a subset of Σ_S denoting a *complete* theory or model (using CWA).
- A planning task (or planning instance) is a 4-tuple $\Pi = \langle S, \mathbf{O}, \mathbf{I}, \mathbf{G} \rangle$, where
 - O is a set of operator (or action types)
 - $\mathbf{I} \subseteq \Sigma_{\mathcal{S}}$ is the initial state
 - ullet $\mathbf{G}\subseteq\Sigma_{\mathcal{S}}$ is the goal specification
- No domain constraints (although present in original formalism)

Operators, Actions & State Change

Operator:



with $\underline{para} \subseteq V$ $\underline{pre} \subseteq \Sigma_{\mathcal{S},V}$ $\underline{eff} \subseteq \Sigma_{\mathcal{S},V} \cup \neg \Sigma_{\mathcal{S},V}$ (element-wise negation) and all variables in \underline{pre} and \underline{eff} are listed in \underline{para} . Also: $\underline{pre}(o)$, $\underline{eff}(o)$.

eff = positive effect literalseff = positive effect literals

- Operator instance or action: Operator with empty parameter list (instantiated schema!)
- State change induced by action:

$$\overbrace{App(S,o)} = \left\{ \begin{array}{ll} S \cup \underline{\mathit{eff}^+(o)} - \neg \underline{\mathit{eff}^-(o)} & \text{if} \underbrace{\mathit{pre}(o)} \subseteq S \& \\ & \mathit{eff}(o) \text{ is cons.} \\ \text{undefined} & \text{otherwise} \end{array} \right.$$

Example Formalization: Logistics

- Logical atoms: at(O,L), in(O,V), airconn(L1,L2), street(L1,L2), plane(V), truck(V)
- Load into truck: load

Parameter list: (O, V, L)

Precondition: at(O,L), at(V,L), truck(V)

Effects: $\sqrt{at(O,L)} in(O,V)$

• Drive operation: drive

Parameter list: (V, L1, L2)

Precondition: at(V, L1), truck(V), street(L1, L2)

Effects: $\neg at(V, L1), at(V, L2)$

- ..
- Some constant symbols: v1, s, t with truck(v1) and street(s, t)
- Action: drive(v1, s, t)

Plans & Successful Executions

- A plan Δ is a sequence of actions
- State resulting from executing a plan:

$$\begin{array}{rcl} Res(S,\langle\rangle) & = & S \\ Res(S,(o;\Delta)) & = & \left\{ \begin{array}{ll} Res(App(S,o),\Delta) & \text{ if } App(S,o) \\ & \text{ is defined} \\ \text{ undefined} & \text{ otherwise} \end{array} \right. \end{array}$$

• Plan Δ is successful or solves a planning task if $Res(\mathbf{I}, \Delta)$ is defined and $\mathbf{G} \subseteq Res(\mathbf{I}, \Delta)$.

A Small Logistics Example

Initial state:
$$S = \left\{ \underbrace{at(p1,c),at(p2,s),at(t1,c),}_{at(t2,c),street(c,s),street(s,c)} \right\}$$

Goal: G =
$$\{at(p1, s), at(p2, c)\}$$

Successful plan:
$$\Delta = \langle load(p), t1, c \rangle, drive(t1, c, s), \underline{unload(p1, t1, s), load(p2, t1, s),} drive(t1, s, c), unload(p2, t1, c) \rangle$$

Other successful plans are, of course, possible

Simplifications: DATALOG- and Propositional STRIPS

- STRIPS as described above allows for unrestricted first-order terms,
 i.e., arbitrarily nested function terms
- \rightarrow Infinite state space
- Simplification: No function terms (only 0-ary = constants)
- → DATALOG-STRIPS
- Simplification: No variables in operators (= actions)
- → Propositional STRIPS
 - Propositional STRIPS used in planning algorithms nowadays (but specification is done using DATALOG-STRIPS)

Beyond STRIPS

Even when keeping all the restrictions of classical planning, one can think of a number of extensions of the planning language.

- General logical formulas as preconditions: Allow all Boolean connectors and quantification
- Conditional effects: Effects that happen only if some additional conditions are true. For example, when pressing the accelerator pedal, the effects depends on which gear has been selected (no, reverse, forward).
- Multi-valued state variables: Instead of 2-valued Boolean variables, multi-valued variables could be used
- Numerical resources: Resources (such as fuel or time) can be effected and be used in preconditions
- Durative actions: Actions can have duration and can be executed concurrently
- Axioms/Constraints: The domain is not only described by operators, but also by additional laws

PDDL: The Planning Domain Description Language

- Since 1998, there exists a bi-annual scientific competition for action planning systems.
- In order to have a common language for this competition, PDDL has been created (originally by Drew McDermott)
- Meanwhile, version 3.1 (IPC-2011) with most of the features mentioned – and many sub-dialects and extensions.
- Sort of "standard" by now.

PDDL Logistics Example

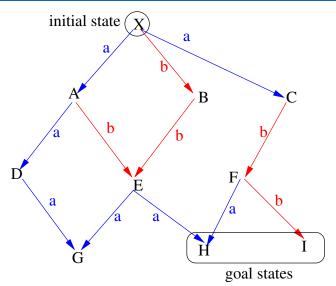
```
(define (domain logistics)
   (:types truck airplane - vehicle
        package vehicle - physobj
        airport location - place
        city place physobj - object)
   (:predicates (in-city ?loc - place ?city - city)
        (at ?obj - physobj ?loc - place)
        (in ?pkg - package ?veh - vehicle))
:action LOAD-TRUCK
   :parameters (?pkg - package ?truck - truck ?loc - place)
   :precondition (and (at ?truck ?loc) (at ?pkg ?loc))
   :effect
                  (and (not (at ?pkg ?loc)) (in ?pkg ?truck)))
                  . . . )
```

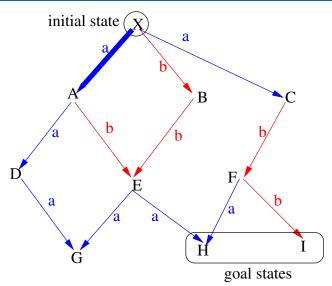
Lecture Overview

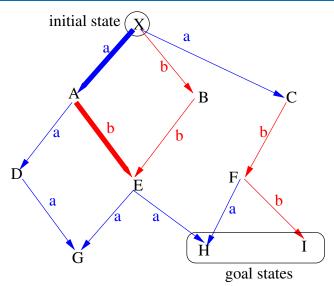
- What is Action Planning?
- Planning Formalisms
- Basic Planning Algorithms
- Computational Complexity
- Current Algorithmic Approaches
- Current Trends in Planning
- Summary

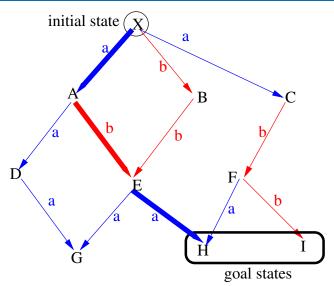
Planning Problems as Transition Systems

- We can view planning problems as searching for goal nodes in a large labeled graph (transition system)
- Nodes are defined by the value assignment to the fluents = states
- Labeled edges are defined by actions that change the appropriate fluents
- Use graph search techniques to find a (shortest) path in this graph!
- Note: The graph can become huge: 50 Boolean variables lead to 2^{50} = 10^{15} states
- → Create the transition system on the fly and visit only the parts that are necessary









Progression Planning: Forward Search

Search through transition system starting at initial state

- Initialize partial plan $\Delta := \bigcirc$ and start at the unique initial state I and make it the current state S
- **2** Test whether we have reached a goal state already: $G \subseteq S$? If so, return plan Δ .
- Select one applicable action oinon-deterministically and
 - compute successor state $S := App(S, o_i)$
 - extend plan $\Delta := \langle \Delta, o_i \rangle$, and continue with step 2.

Instead of non-deterministic choice use some search strategy. Progression planning can be easily extended to more expressive

planning languages Waly leaden.

$$\mathcal{S} = \{a, b, c, d\},$$

$$\mathbf{O} = \{o_1 = \langle \emptyset, \{a, b\}, \{\neg b, c\} \rangle, \\ o_2 = \langle \emptyset, \{a, b\}, \{\neg a, \neg b, d\} \rangle, \\ o_3 = \langle \emptyset, \{c\}, \{b, d\} \rangle,$$

$$\mathbf{I} = \{a, b\}$$

$$\mathbf{G} = \{b, d\}$$



$$S = \{a, b, c, d\},\$$

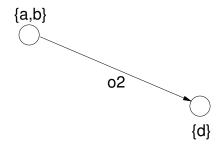
$$O = \{o_1 = \langle \emptyset, \{a, b\}, \{\neg b, c\} \rangle,\$$

$$o_2 = \langle \emptyset, \{a, b\}, \{\neg a, \neg b, d\} \rangle,\$$

$$o_3 = \langle \emptyset, \{c\}, \{b, d\} \rangle,\$$

$$I = \{a, b\}\$$

$$G = \{b, d\}$$



$$S = \{a, b, c, d\},\$$

$$O = \{o_1 = \langle \emptyset, \{a, b\}, \{\neg b, c\} \rangle,\$$

$$o_2 = \langle \emptyset, \{a, b\}, \{\neg a, \neg b, d\} \rangle,\$$

$$o_3 = \langle \emptyset, \{c\}, \{b, d\} \rangle,\$$

$$I = \{a, b\}$$

$$G = \{b, d\}$$

$$\{a, b\}$$

$$o_1 \qquad \{a, c\}$$

$$o_2 \qquad \{d\}$$

$$S = \{a, b, c, d\},\$$

$$O = \{o_1 = \langle \emptyset, \{a, b\}, \{\neg b, c\} \rangle,\$$

$$o_2 = \langle \emptyset, \{a, b\}, \{\neg a, \neg b, d\} \rangle,\$$

$$o_3 = \langle \emptyset, \{c\}, \{b, d\} \rangle,\$$

$$I = \{a, b\}\$$

$$G = \{b, d\}$$

$$\{a,b\}$$

$$o_1 \qquad \{a,c\} \qquad o_3 \qquad \{a,b,c,d\}$$

$$G = \{b,d\}$$

$$G = \{b,d\}$$

Regression Planning: Backward Search

Search through transition system starting at goal states. Consider sets of states, which are described by the atoms that are necessarily true in them

- Initialize partial plan $\Delta := \langle \ \rangle$ and set $\mathbf{S} := \mathbf{G}$
- Test whether we have reached the unique initial state already: $I \supseteq S$? If so, return plan Δ .
- Select one action o_i non-deterministically which does not make (sub-)goals false (S $\cap \neg eff^-(o_i) = \emptyset$) and
 - compute the regression of the description S through o_i :

$$\mathbf{S} := \mathbf{S} - eff^{+}(o_{i}) \cup pre(o_{i})$$

• extend plan $\Delta := \langle o_i, \Delta \rangle$ and continue with step 2.

Instead of non-deterministic choice use some search strategy
Regression becomes much more complicated, if e.g. conditional
effects are allowed. Then the result of a regression can be a general
Boolean formula

$$S = \{a, b, c, d, e\},$$

$$\mathbf{O} = \{o_1 = \langle \emptyset, \{b\}, \{\neg b, c\} \rangle, o_2 = \langle \emptyset, \{e\}, \{b\} \rangle, o_3 = \langle \emptyset, \{c\}, \{b, d, \neg e\} \rangle,$$

$$\mathbf{I} = \{a, b\}$$

$$\mathbf{G} = \{b, d\}$$

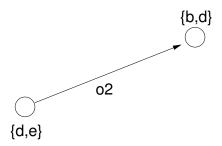


$$S = \{a, b, c, d, e\},$$

$$\mathbf{O} = \{o_1 = \langle \emptyset, \{b\}, \{\neg b, c\} \rangle, o_2 = \langle \emptyset, \{e\}, \{b\} \rangle, o_3 = \langle \emptyset, \{c\}, \{b, d, \neg e\} \rangle,$$

$$\mathbf{I} = \{a, b\}$$

$$\mathbf{G} = \{b, d\}$$



$$S = \{a, b, c, d, e\},$$

$$O = \{o_1 = \langle \emptyset, \{b\}, \{\neg b, c\} \rangle, o_2 = \langle \emptyset, \{e\}, \{b\} \rangle, o_3 = \langle \emptyset, \{c\}, \{b, d, \neg e\} \rangle,$$

$$I = \{a, b\}$$

$$G = \{b, d\}$$

$$\{c\}$$

$$03$$

$$\{b, d\}$$

{d,e}

$$S = \{a, b, c, d, e\},\$$

$$O = \{o_1 = \langle \emptyset, \{b\}, \{\neg b, c\} \rangle,\$$

$$o_2 = \langle \emptyset, \{e\}, \{b\} \rangle,\$$

$$o_3 = \langle \emptyset, \{c\}, \{b, d, \neg e\} \rangle,\$$

$$I = \{a, b\}$$

$$G = \{b, d\}$$

$$\{b\}$$

$$01 \qquad \{c\}$$

$$03 \qquad \{b, d\}$$

$$|=\{a, b\}$$

$$\{d, e\}$$

Other Types of Search

- Of course, other types of search are possible.
- Change perspective: Do not consider the transition system as the space we have to explore, but consider the search through the space of (incomplete) plans:
 - Progression search: Search through the space of plan prefixes
 - Regression search: Search through plan suffixes
- Partial order planning:
 - Search through partially ordered plans by starting with the empty plan and trying to satisfy (sub-)goals by introducing new actions (or using old ones)
 - Make ordering choices only when necessary to resolve conflicts

- What is Action Planning?
- Planning Formalisms
- Basic Planning Algorithms
- Computational Complexity
- Current Algorithmic Approaches
- Current Trends in Planning
- Summary

The Planning Problem - Formally

Definition (Plan existence problem (PLANEX))

Instance: $\Pi = \langle \mathcal{S}, \mathbf{O}, \mathbf{I}, \mathbf{G} \rangle$.

Question: Does there exist a plan Δ that solves Π , i.e., $Res(\mathbf{I}, \Delta) \supseteq \mathbf{G}$?

Definition (Bounded plan existence problem (PLANLEN))

Instance: $\Pi = \langle S, \mathbf{O}, \mathbf{I}, \mathbf{G} \rangle$ and a positive integer n.

Question: Does there exist a plan Δ of length n or less that solves Π ?

From a practical point of view, also PLANGEN (generating a plan that solves Π) and PLANLENGEN (generating a plan of length n that solves Π) and PLANOPT (generating an optimal plan) are interesting (but at least as hard as the decision problems).

June 7, 2019

29/42

Basic STRIPS with First-Order Terms

- The state space for STRIPS with general first-order terms is infinite
- We can use function terms to describe (the index of) tape cells of a Turing machine
- We can use operators to describe the Turing machine control
- The existence of a plan is then equivalent to the existence of a successful computation on the Turing machine
- PLANEX for STRIPS with first-order terms can be used to decide the Halting problem

Theorem

PLANEX for STRIPS with first-order terms is undecidable.

Propositional STRIPS

Theorem

PLANEX is **PSPACE-complete** for propositional STRIPS.

- → Membership follows because we can successively guess operators and compute the resulting states (needs only polynomial space)
- → Hardness follows using again a generic reduction from TM acceptance. Instantiate polynomially many tape cells with no possibility to extend the tape (only poly. space, can all be generated in poly. time)
 - PLANLEN is also PSPACE-complete (membership is easy, hardness follows by setting $k=2^{|\Sigma|}$)

Restrictions on Plans

- If we restrict the length of the plans to be short, i.e., only polynomial in the size of the planning task, PLANEX becomes NP-complete
- Similarly, if we use a unary representation of the natural number k, then PLANLEN becomes NP-complete
- → Membership obvious (guess & check)
 - Hardness by a straightforward reduction from SAT or by a generic reduction.
 - One source of complexity in planning stems from the fact that plans can become very long
 - We are only interested in short plans!
 - We can use methods for NP-complete problems if we are only looking for "short" plans.

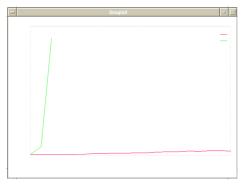
- What is Action Planning?
- Planning Formalisms
- Basic Planning Algorithms
- Computational Complexity
- Current Algorithmic Approaches
- Current Trends in Planning
- Summary

Current Approaches

- Planning as satisfiability: Iterative deepening.
- Planning with answer set programming.
- Symbolic planning with BDDs (finding many plans or non-det. plans)
- Heuristic forward-search planning (HSP, FF, FD)

Heuristic Search Planning

- Use an automatically generated heuristic estimator in order to select the next action or state
- Depending on the search scheme and the heuristic, the plan might not be the shortest one
- → It is often easier to go for sub-optimal solutions (remember Logistics)



Runtime of Heuristic search planner (red) ys. iterative deepening (green) on Gripper

Deriving Heuristics: Relaxations

- General principle for deriving heuristics:
 - Define a simplification (relaxation) of the problem and take the difficulty of a solution for the simplified problem as an heuristic estimator
- Example: straight-line distance on a map to estimate the travel distance
- Example: decomposition of a problem, where the components are solved ignoring the interactions between the components, which may incur additional costs
- In planning, one possibility is to ignore negative effects

Ignoring Negative Effects: Example

- In Logistics: The negative effects in load and drive are ignored:
- Simplified load operation: load(O, V, P)Precondition: at(O, P), at(V, P), truck(V)Effects: at(O, P), in(O, V)
- After loading, the package is still at the place and also inside the truck
- Simplified drive operation: drive(V, P1, P2)Precondition: at(V, P1), truck(V), street(P1, P2)Effects: at(V, P1), at(V, P2)
- After driving, the truck is in two places!
- \rightarrow We want the length of the shortest relaxed plan $\rightsquigarrow h^+(s)$
 - How difficult is monotonic planning?

- What is Action Planning?
- Planning Formalisms
- Basic Planning Algorithms
- Computational Complexity
- Current Algorithmic Approaches
- Current Trends in Planning
- Summary

Current Trends in Al Planning

- Developing and analyzing heuristics
- Developing and analyzing pruning techniques
- Develping new search techniques
- Extending the expressiveness of planning formalisms (and extending planning algorithms) in order to deal with
 - temporal planning,
 - planning with non-deterministic actions,
 - planning under partial observability,
 - planning with probabilistic effects,
 - multi-agent planning,
 - planning with epistemic goals,
- Reasoning about plans, e.g., diagnosing failures
- Judging morality of plans
- Applying/integrating planning technology
- Learning and planning
- . . .

39/42

Our Interests

- Foundation / theory
- Extending planning technology in order to cope with multi-agent scenarios and epistemic goals
- Using EVMDDs in modelling state-dependent costs
- Using planning techniques and extending them for robot control
- Using planning methodolgy in application in general
- Exploring the ethical dimension of planning systems

- What is Action Planning?
- Planning Formalisms
- Basic Planning Algorithms
- Computational Complexity
- Current Algorithmic Approaches
- Current Trends in Planning
- Summary

Summary

- Rational agents need to plan their course of action
- In order to describe planning tasks in a domain-independent, declarative way, one needs planning formalisms
- Basic STRIPS is a simple planning formalism, where actions are described by their preconditions in form of a conjunction of atoms and the effects are described by a list of literals that become true and false
- PDDL is the current "standard language" that has been developed in connection with the international planning competition
- Basic planning algorithms search through the space created by the transition system or through the plan space.
- Planning with STRIPS using first-order terms is undecidable
- Planning with propositional STRIPS is PSPACE-complete
- Since 1992, we have reasonably efficient planning method for propositional, classical STRIPS planning
- You can learn more about it in our planning class next term.