

Introduction to Artificial Intelligence

Planning

Ing. Tomas Borovicka

Department of Theoretical Computer Science (KTI), Faculty of Information Technology (FIT)
Czech Technical University in Prague (CVUT)

BIE-ZUM, LS 2013/14, 6. lecture



<https://edux.fit.cvut.cz/courses/BIE-ZUM/>

Summary of Previous Lecture

- Evolutionary Computation

- ▶ Family of global meta-heuristic or stochastic optimization methods.
- ▶ Algorithms typically imitate some principle of natural evolution as method to solve optimization problems.
- ▶ Genetic Algorithm
 - ★ Universal "black box" solver for optimization using binary strings.
- ▶ Genetic Programming
 - ★ Approach for general automatic programming.
 - ★ Originally used to evolve Lisp programs.
- ▶ Evolutionary Programming
 - ★ Motivation was to generate alternative approach to artificial intelligence.
 - ★ Early versions applied to the evolution of transition table of finite state machines.
- ▶ Evolution Strategies
 - ★ Based on the concept of the "evolution of evolution".

General Problem Solving

- Traditional approach to solve a problem:

Problem \longrightarrow Human \longrightarrow **Design of special algorithm** \longrightarrow Solution

- Disadvantages:

- ▶ design and implementation of specialized algorithm is expensive,
- ▶ optimization and tuning of the algorithm is time consuming.

- **General Problem Solving:**

Problem \longrightarrow Human \longrightarrow Problem formalization \longrightarrow **Solver** \longrightarrow Solution

- ▶ The user must only formalize the problem.

General Problem Solving

General principle:

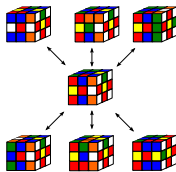
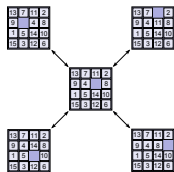
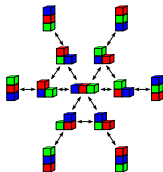
Problem \longrightarrow Model \longrightarrow Language \longrightarrow Solver \longrightarrow Solution

- 1 Having a **problem** P ,
- 2 choose a **model** M ,
- 3 choose a **language** L for definition of model M ,
- 4 choose a **solver** S for solving problems in L ,
- 5 leave solver **automatically solve** $S(L(P))$,
- 6 **interpret** $S(L(P))$ in the context of P .

Automatic Planning

- Sliding puzzles,
- Rubik's cube,
- Blocks world,

All these problems can be simply defined using state space, where each state represents some **structured configuration** and actions make changes in this configuration.



Planning vs Search

Planning problem is essentially search problem. However there are differences that make reasonable to treat them in a different way.

	Search	Planning
States	data structures	logical sentences
Actions	code	preconditions / effects
Goal	code	logical sentences
Plan	sequence of actions	constrains on actions

The main benefits are:

- Unified action and goal representation (logical language).
- Independent solution of sub-goals .
- Flexible construction of solution, relax requirement for sequential construction.

Classical Planning

Assumptions are:

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

Planning Problem

Objective is to get tea, biscuits and a book.

- Initial state:

The agent is at *home* without tea, biscuits, book.

- Goal state:

The agent is at *home* with tea, biscuits, book.

States

States can be represented by predicates such as:

- $At(x)$ - agent is at a specific position x .
- $Have(y)$ - agent has an item y .
- $Sells(x,y)$ - some location x sells item y .

Actions

- $\text{Go}(y)$ - agent goes to y ,
causes $\text{At}(y)$ to be true.
- $\text{Buy}(z)$ - agent buys z ,
causes $\text{Have}(z)$ to be true.
- $\text{Steal}(z)$ - agent steals z ,
causes $\text{Have}(z)$ to be true.

Problem Representation

- Initially attempted to represent planning problems through variants of predicate calculus, like first order logic and propositional calculus.
- Initial State:** We are home we do not have tea, we do not have biscuits and we do not have a book.

$$At(Home, s_0) \wedge \neg Have(Tea, s_0) \wedge \neg Have(Biscuits, s_0) \wedge \neg Have(Book, s_0)$$

- Goal State:** There exists some state where we are at home, we have a tea, we have biscuits and we have a book.

$$\exists s At(Home, s) \wedge Have(Tea, s) \wedge Have(Biscuits, s) \wedge Have(Book, s)$$

Operators

$$\forall a, s \text{ Have}(Tea, \text{Result}(a, s)) \iff [(a = \text{Buy}(Tea) \wedge \text{At}(TeaShop, s)) \\ \wedge (\text{Have}(Tea, s) \wedge a \neq \text{Drop}(Tea))]$$

- $\text{Result}(a, s)$ names the situation resulting from executing the action a in the situation s .
- $\text{Drop}(z)$ - agent drops z ,
 - ▶ causes $\text{Have}(z)$ to be false.

The frame problem

- We have to write rules for all things that does not change.
- Everything that just will not change, we have to explicitly specify in predicate logic to say that it will not change.
- Makes representation of a problem very complex.

Resolution: If very few things are changing at a time, then, it is always easier to model it as changes rather than anything else.

STRIPS

- STanford Research Institute Problem Solver

A STRIPS instance is composed of:

- An initial state.
- The goal states.
- The set of actions (operators) described by
 - ▶ preconditions - must be satisfied before the action is performed.
 - ▶ effects - established after the action is performed.

STRIPS

STRIPS

STRIPS instance is a quadruple (P, A, \mathcal{I}, G) , where

- P is a set of conditions,
- A is a set of actions (operators) described by $\text{precond}(a) \subseteq P$ and $\text{effect}(a) \subseteq P$ (in original version $\text{add}(a) \subseteq P, \text{del}(a) \subseteq P$),
- $\mathcal{I} \subseteq P$ is **initial state**,
- $G \subseteq P$ is **goal state**.

Planning problem \neq state space.

STRIPS: Block World Example 1



$$P = \{ \textit{red-on-ground}, \textit{red-on-top}, \textit{red-on-green}, \textit{red-on-blue}, \\ \textit{green-on-ground}, \textit{green-on-top}, \textit{green-on-red}, \textit{green-on-blue}, \\ \textit{blue-on-ground}, \textit{blue-on-top}, \textit{blue-on-red}, \textit{blue-on-green} \}$$

$$A = \{ \textit{move-red-ground-green}, \textit{move-red-ground-blue}, \textit{move-red-green-blue}, \\ \textit{move-red-blue-green}, \textit{move-red-green-ground}, \textit{move-red-blue-ground}, \\ \textit{move-green-ground-red}, \textit{move-green-ground-blue}, \dots \}$$

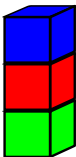
$$\mathcal{I} = \{ \textit{red-on-green}, \textit{green-on-ground}, \textit{blue-on-top}, \textit{blue-on-red} \}$$

$$G = \{ \textit{green-on-red}, \textit{red-on-ground}, \textit{blue-on-top}, \textit{blue-on-green} \} \\ \textit{move-blue-ground-green})$$

STRIPS: Block World Example 1

```

red-on-ground = 0
red-on-top    = 0
red-on-green  = 1
red-on-blue   = 0
blue-on-ground = 0
blue-on-top   = 1
blue-on-red   = 1
blue-on-green = 0
green-on-ground = 1
green-on-top  = 0
green-on-red  = 0
green-on-blue = 0
  
```

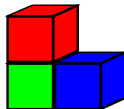


```

move-blue-red-ground
pre = { blue-on-top, blue-on-red }
add = { blue-on-ground,
        red-on-top }
del = { blue-on-red }
  
```

```

move-blue-ground-red
pre = { blue-on-ground,
        blue-on-top,
        red-on-top }
add = { blue-on-red }
del = { blue-on-ground,
        red-on-top }
  
```



```

red-on-ground = 0
red-on-top    = 1
red-on-green  = 1
red-on-blue   = 0
blue-on-ground = 1
blue-on-top   = 1
blue-on-red   = 1
blue-on-green = 0
green-on-ground = 1
green-on-top  = 0
green-on-red  = 0
green-on-blue = 0
  
```

Block World Example 2



- **Objects** : $U = \{R, G, B\}$,
- **Predicates**: $P = \{on, on-ground, on-top, distinct\}$,

Problem in STRIPS (P', A, \mathcal{I}, G):

$$P' = \{on(x, y), on-ground(x), on-top(x), distinct(x, y) \mid x, y \in \{R, G, B\}\},$$

$$A = \{move(what, from, to), from-ground(what, to), to-ground(what, from)\}$$

$$\mathcal{I} = \{on-ground(G), on(G, R), on(R, B), on-top(B)\}$$

$$\cup \{distinct(x, y) \mid x, y \in \{R, G, B\}, x \neq y\}$$

$$G = \{on-ground(R), on(R, G), on(G, B), on-top(B)\},$$

STRIPS: Block World Example 2



For each action we define preconditions and effects (add and del):

- $move(what, from, to)$
 - ▶ $pre(move) = \{on(from, what), on-top(what), on-top(to), distinct(what, to)\}$,
 - ▶ $add(move) = \{on(to, what), on-top(from)\}$,
 - ▶ $del(move) = \{on(from, what), on-top(to)\}$,
- $from-ground(what, to)$
 - ▶ $pre(from-ground) = \{on-ground(what), on-top(what), on-top(to)\}$
 - ▶ $add(from-ground) = \{on(to, what)\}$
 - ▶ $del(from-ground) = \{on-ground(what), on-top(to)\}$
- $to-ground(what, from)$
 - ▶ $pre(to-ground) = \{on(from, what), on-top(what)\}$
 - ▶ $add(to-ground) = \{on-ground(what), on-top(from)\}$
 - ▶ $del(to-ground) = \{on(from, what)\}$

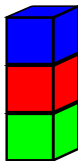
STRIPS: Block World Example 2

```

on-ground(green)
on(green,red)
on(red,blue)
on-top(blue)

distinct(red,green)
distinct(green,red)
...

```



```

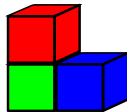
to-ground(what,from)
[what:=blue,from:=red]
pre = { on(from,what),
        on-top(what) }
add = { on-ground(what),
        on-top(from) }
del = { on(from,what) }

```

```

from-ground(what,to)
[what:=blue,to:=red]
pre = { on-ground(what),
        on-top(what),
        on-top(to) }
add = { on(to,what) }
del = { on-ground(what),
        on-top(to) }

```



```

on-ground(green)
on(green,red)
on-top(blue)
on-ground(blue)
on-top(red)

distinct(red,green)
distinct(green,red)
...

```

Representing States

- States are represented by positive function-free literals (atoms).

Initial:

$$At(Home) \wedge Sells(BS, Book) \wedge Sells(TS, Tea) \wedge Sells(TS, Biscuits)$$

Goal:

$$At(Home) \wedge Have(Tea) \wedge Have(Biscuits) \wedge Have(Book)$$

- Closed World: unmentioned literals are false.
- In later definition states can also contain variables

$$At(x) \wedge Sells(x, Tea).$$

Representing Actions

- **Action description** - serves as a name.
- **Precondition** - a conjunction of positive literals.
- **Effect** - a conjunction of positive and negative literals.
 - ▶ The original version had an *add* list and *del* list
(effect of $P \wedge \neg Q$ means add P , delete Q)

$Op(ACTION : Go(There),$
 $PRECOND : At(Here) \wedge Path(Here, there),$
 $EFFECT : At(there) \wedge \neg At(Here))$

Representing Plans

- A set of steps, where each step is one of the operators of the problem.
- A set of step ordering constraints. Each ordering constraint is of the form $S_i \prec S_j$ indicating S_i must occur sometime before S_j
- A set of variable binding constraints of the form $v = x$ where v is a variable in some step, and x is either a constant or another variable.
- A set of causal links written as $S \xrightarrow{c} S'$ indicating S satisfies the precondition c for S' .

Example

Actions:

*Op(ACTION : RightShoe,
PRECONDITION : RightSockOn,
EFFECT : RightShoeOn)*

*Op(ACTION : RightSock,
EFFECT : RightShoeOn)*

*Op(ACTION : LeftShoe,
PRECONDITION : LeftSockOn,
EFFECT : LeftShoeOn)*

*Op(ACTION : LeftShoe,
EFFECT : LeftShoeOn)*

Example: Initial Plan

```
Plan(  
  STEPS : {  
    S1 : Op(ACTION : start),  
    S2 : Op(ACTION : finish,  
            PRECOND : RightShoeOn  $\wedge$  LeftShoeOn)},  
  ODRERINGS : {S1  $\prec$  S2},  
  BINDINGS : {},  
  LINKS : {})
```

Partial Order Planning

- We add steps from the given set of actions in order to satisfy not achieved preconditions.
- We finish when all preconditions of each step has been satisfied.
- Any topologically ordered sequence of actions is solution.

Example

- Initial State:

$Op(ACTION : Start,$
 $EFFECT : At(Home) \wedge Sells(BS, Book)$
 $\wedge Sells(TS, Tea) \wedge Sells(TS, Biscuits))$

- Goal:

$Op(ACTION : Finish,$
 $PRECOND : At(Home) \wedge Have(Tea)$
 $\wedge Have(Biscuits) \wedge Have(Book)$

Example

- Actions

$Op(ACTION : Go(There),$
 $PRECOND : At(Here),$
 $EFFECT : At(there) \wedge \neg At(Here))$

$Op(ACTION : Buy(x),$
 $PRECOND : At(store) \wedge Sells(store, x),$
 $EFFECT : Have(x))$

Partial Order Planning: Pseudo-code Sketch

Algorithm 1 Partial Order Planning (POP)

```
1:  $plan \leftarrow \text{INIT\_MINIMAL\_PLAN}(initial, goal)$ 
2: loop
3:   if  $\text{SOLUTION}(plan)$  then return  $plan$ 
4:   end if
5:    $S_{need}, c \leftarrow \text{SELECT\_SUBGOAL}(plan)$ 
6:    $\text{CHOOSE\_OPERATOR}(plan, OPERATORS, S_{need}, c)$ 
7:    $\text{RESOLVE\_THREATS}(plan)$ 
8: end loop
```

Algorithm 2 Partial Order Planning (POP)

```
1: function  $\text{SELECT\_SUBGOAL}(plan)$ 
2:   pick a step  $S_{need} \in STEPS$  with a precondition  $c$  that has not been achieved
3:   return  $S_{need}, c$ 
4: end function
```

Algorithm 3 Partial Order Planning (POP)

```
1: function CHOOSE_OPERATOR(plan, OPERATORS,  $S_{need}$ , c)
2:   choose s step  $S_{add} \in OPERATORS \cup STEPS$  that has c as an effect
3:   if  $S_{add} = \{\}$  then FAIL
4:   end if
5:   add the causal link  $S_{add} \xrightarrow{c} S_{need}$  to LINKS
6:   add the ordering constrain  $S_{add} \prec S_{need}$  to ORDERINGS
7:   if  $S_{add}$  is newly added step from operators then
8:     add  $S_{add}$  to STEPS
9:     add  $S_{start} \prec S_{add} \prec S_{goal}$  to ORDERINGS
10:  end if
11: end function
```

Algorithm 4 Partial Order Planning (POP)

```
1: function RESOLVE_THREATS(plan)
2:   for all  $S_{threat}$  that threatens a link  $S_i \xrightarrow{c} S_j \in \text{LINKS}$  do
3:     Promote: add  $S_{threat} \prec S_i$  or Demote: add  $S_j \prec S_{threat}$ 
4:     if  $\neg \text{CONSISTENT}(\text{plan})$  then FAIL
5:     end if
6:   end for
7: end function
```

Partially Instantiated Operators

- So far we have not mentioned anything about binding constraints.
- Should an operator that has the effect, say, $\neg At(x)$, be considered a threat to the condition $At(Home)$?

Dealing with possible threats:

- Resolve with an equality constraint,
 - ▶ bind x with something that resolves the threat (e.g. $x = TS$)
- Resolve with an inequality constrain.
 - ▶ Add constrain that x can not be bound to $Home$.
- Resolve later.
 - ▶ Ignore possible threats. If $x = Home$ is added later into the plan, try to resolve by promotion or demotion.

Algorithm 5 Partial Order Planning (POP)

```

1: function CHOOSE_OPERATOR(plan, OPERATORS,  $S_{need}$ , c)
2:   choose s step  $S_{add} \in OPERATORS \cup STEPS$  that has c as an effect such
   that  $u = UNIFY(c, c', BINDINGS)$ 
3:   if  $S_{add} = \{\}$  then FAIL
4:   end if
5:   add u to BINDINGS
6:   add the causal link  $S_{add} \xrightarrow{c} S_{need}$  to LINKS
7:   add the ordering constrain  $S_{add} \prec S_{need}$  to ORDERINGS
8:   if  $S_{add}$  is newly added step from operators then
9:     add  $S_{add}$  to STEPS
10:    add  $S_{start} \prec S_{add} \prec S_{goal}$  to ORDERINGS
11:   end if
12: end function

```
