



Planning

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2.04.2009



Plan for today

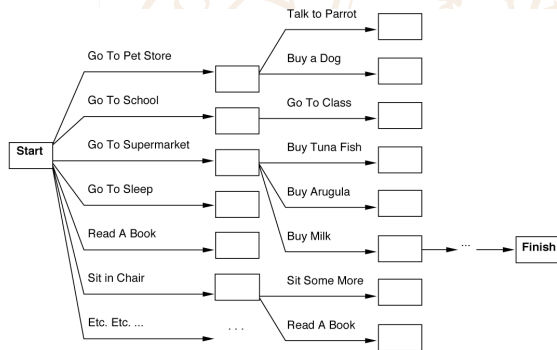
- Search vs. planning
- STRIPS operators, ADL, PDDL
- Partial-order planning
- Graphplan



Search vs. planning

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

Standard search algorithms seem to fail miserably:



After-the-fact heuristic/goal test inadequate



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

Planning in situation calculus

$PlanResult(p, s)$ is the situation resulting from executing p in s

$PlanResult([], s) = s$

$PlanResult([a|p], s) = PlanResult(p, Result(a, s))$

Initial state $At(Home, S_0) \wedge \neg Have(Milk, S_0) \wedge \dots$

Actions as Successor State axioms

$Have(Milk, Result(a, s)) \Leftrightarrow$

$[(a = Buy(Milk) \wedge At(Supermarket, s)) \vee (Have(Milk, s) \wedge a \neq \dots)]$

Query:

$s = PlanResult(p, S_0) \wedge At(Home, s) \wedge Have(Milk, s) \wedge \dots$

Solution:

$p =$

$[Go(Supermarket), Buy(Milk), Buy(Bananas), Go(HWS), \dots]$

Principal difficulty: unconstrained branching, hard to apply heuristics

STRIPS operators

Properly 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

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

$Buy(x)$

$Have(x)$

STRIPS vs ADL

STRIPS - Stanford Reserch Institute Problem Solver

ADL - Action Description Language

PDDL - Planning Domain Description Language

	STRIPS	ADL
State	Positive literals	Positive and negative literals
CWA	Present	Absent
Preconditions	positive literals	Logical sentences
Effects	$P \wedge \neg Q$ add, delete	$P \wedge \neg Q$ add all this
Goals	ground literals	quantified formulae
Goals	conjunction	arbitrary
Effects	conjunction	conditionals allowed

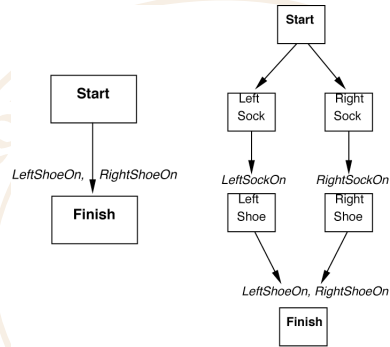
Partially ordered plans

Partially ordered collection of steps with

- *Start step* has the initial state description as its effect
- *Finish step* has the goal description as its precondition
- *causal links* from outcome of one step to precondition of another
- *temporal ordering* between pairs of steps

Definition: *open condition* is a precondition of a step not yet causally linked

Partially ordered plans



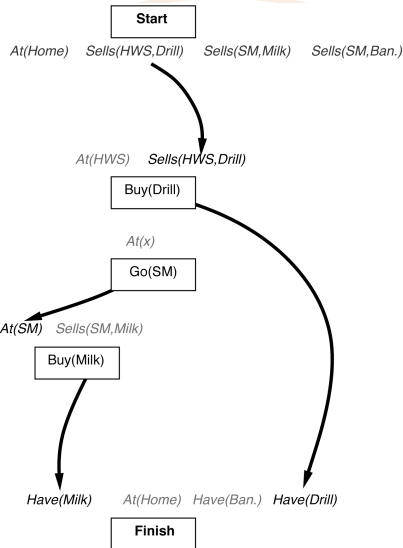
A plan is *complete* iff every precondition is achieved

A precondition is *achieved* iff it is the effect of an earlier step and no *possibly intervening* step undoes it

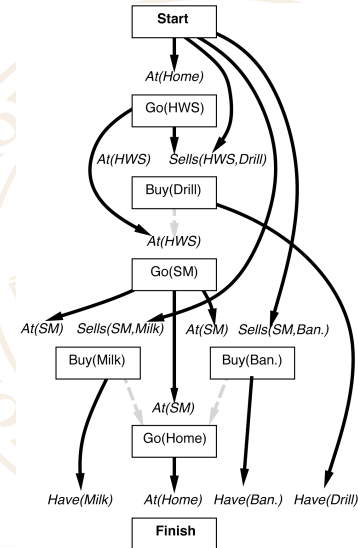
Example



Example



Example



Planning process

Operators on partial plans:
add a link from an existing action to an open condition
add a step to fulfill an open condition
order one step wrt another

Gradually move from incomplete/vague plans to complete, correct plans

Backtrack if an open condition is unachievable or if a conflict is unresolvable

POP algorithm sketch

```
function POP(initial, goal, operators) returns plan
```

```
plan ← MAKE-MINIMAL-PLAN(initial, goal)
```

```
loop do
```

```
  if SOLUTION?(plan) then return plan
```

```
   $S_{need}, c \leftarrow$  SELECT-SUBGOAL(plan)
```

```
  CHOOSE-OPERATOR(plan, operators, Sneed, c)
```

```
  RESOLVE-THREATS(plan)
```

```
end
```

```
function SELECT-SUBGOAL(plan) returns  $S_{need}, c$ 
```

```
pick a plan step  $S_{need}$  from STEPS(plan)
```

```
with a precondition  $c$  that has not been achieved
```

```
return  $S_{need}, c$ 
```

POP algorithm contd.

```
procedure CHOOSE-OPERATOR(plan, operators, Sneed, c)
```

```
choose a step  $S_{add}$  from operators or STEPS(plan) that has  $c$  as an effect
```

```
if there is no such step then fail
```

```
add the causal link  $S_{add} \xrightarrow{c} S_{need}$  to LINKS(plan)
```

```
add the ordering constraint  $S_{add} \prec S_{need}$  to ORDERINGS(plan)
```

```
if  $S_{add}$  is a newly added step from operators then
```

```
  add  $S_{add}$  to STEPS(plan)
```

```
  add  $Start \prec S_{add} \prec Finish$  to ORDERINGS(plan)
```

```
procedure RESOLVE-THREATS(plan)
```

```
for each  $S_{threat}$  that threatens a link  $S_i \xrightarrow{c} S_j$  in LINKS(plan) do
```

```
  choose either
```

```
    Demotion: Add  $S_{threat} \prec S_i$  to ORDERINGS(plan)
```

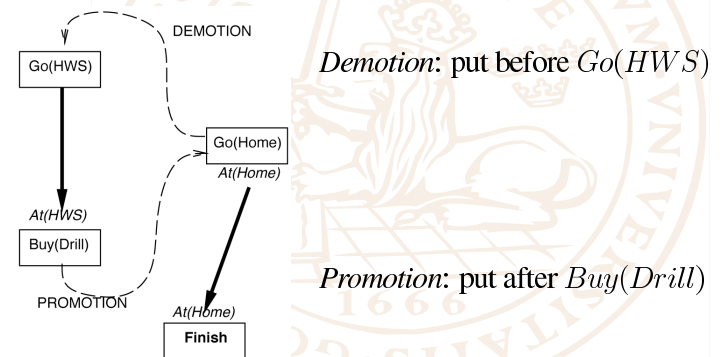
```
    Promotion: Add  $S_j \prec S_{threat}$  to ORDERINGS(plan)
```

```
  if not CONSISTENT(plan) then fail
```

```
end
```

Clobbering and promotion/demotion

A *clobberer* is a potentially intervening step that destroys the condition achieved by a causal link. E.g., $Go(Home)$ clobbers $At(HWS)$:



Properties of POP

Nondeterministic algorithm: backtracks at *choice* points on failure:

- choice of S_{add} to achieve S_{need}
- choice of demotion or promotion for clobberer
- selection of S_{need} is irrevocable

POP is sound, complete, and *systematic* (no repetition)

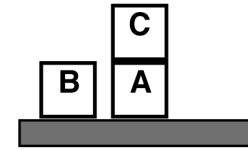
Extensions for disjunction, universals, negation, conditionals

Can be made efficient with good heuristics derived from problem description

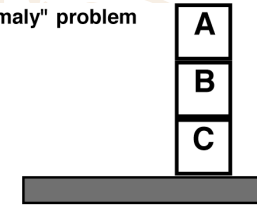
Particularly good for problems with many loosely related subgoals

Example: Blocks world

"Sussman anomaly" problem



Start State



Goal State

$Clear(x) On(x,z) Clear(y)$

PutOn(x,y)

$\sim On(x,z) \sim Clear(y)$
 $Clear(z) On(x,y)$

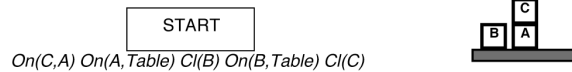
$Clear(x) On(x,z)$

PutOnTable(x)

$\sim On(x,z) Clear(z) On(x, Table)$

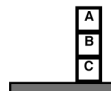
+ several inequality constraints

Example contd.

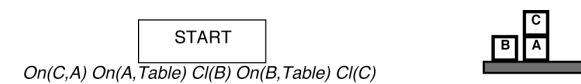


$On(A,B) On(B,C)$

FINISH



Example contd.

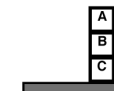


$Cl(B) On(B,z) Cl(C)$

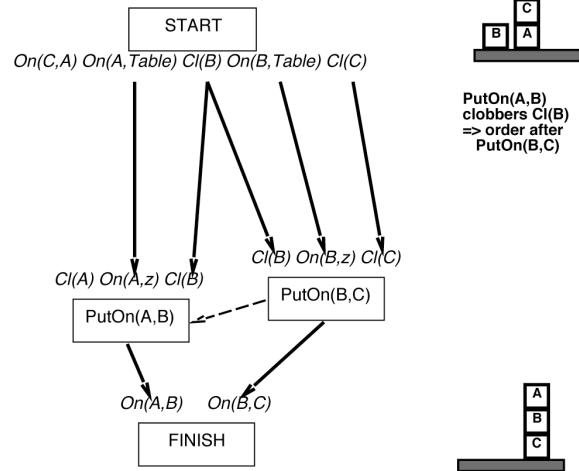
PutOn(B,C)

$On(A,B) On(B,C)$

FINISH



Example contd.



Example contd.

