

Partial-Order Planning

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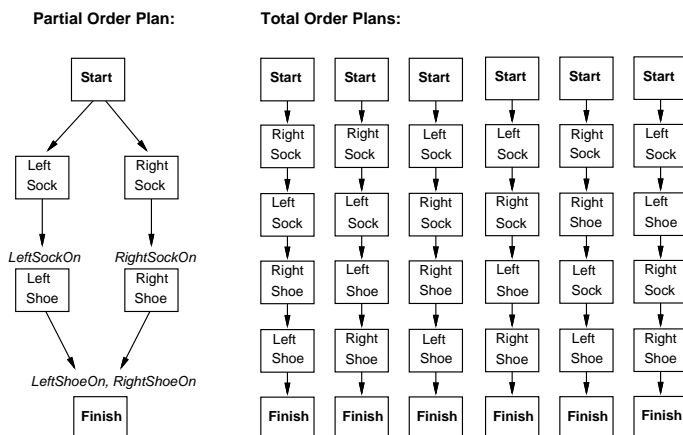
State-Space vs. Plan-Space

- **State-space (situation space)** planning algorithms search through the space of possible states of the world searching for a path that solves the problem.
- They can be based on **progression**: a forward search from the initial state looking for the goal state.
- Or they can be based on **regression**: a backward search from the goals towards the initial state
- STRIPS is an incomplete regression-based algorithm.
- **Plan-space** planners search through the space of partial plans, which are sets of actions that may not be totally ordered.
- **Partial-order** planners are plan-based and only introduce ordering constraints as necessary (**least commitment**) in order to avoid unnecessarily searching through the space of possible orderings.

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Partial Order Plans

- Plan in which not all actions are ordered



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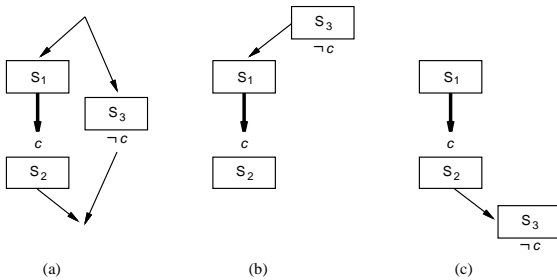
Plans, Causal Links, and Threats

- A *plan* is a three tuple $\langle A, O, L \rangle$
 - A: A set of actions in the plan, $\{A_1, A_2, \dots, A_n\}$
 - O: A set of ordering constraints on actions $\{A_i < A_j, A_k < A_l, \dots, A_m < A_n\}$. These must be *consistent*, i.e. there must be at least one total ordering of actions in A that satisfy all the constraints.
 - B: A set of variable binding constraints $\{v=x, \dots\}$
 - L: a set of causal links showing how actions support each other
- A *causal link*, $A_p \rightarrow^Q A_c$, indicates that action A_p has an effect Q that achieves precondition Q for action A_c .
- A *threat*, is an action A_t that can render a causal link $A_p \rightarrow^Q A_c$ ineffective because:
 - $O \cup \{A_p < A_t < A_c\}$ is consistent
 - A_t has $\neg Q$ as an effect

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Threat Removal

- Threats must be removed to prevent a plan from failing.
- **Demotion** adds the constraint $A_t < A_p$ to prevent clobbering, i.e. push the clobberer before the producer.
- **Promotion** adds the constraint $A_c < A_t$ to prevent clobbering, i.e. push the clobberer after the consumer.



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Initial (Null) Plan

- Initial plan has
 - $A = \{A_0, A_\infty\}$
 - $O = \{A_0 < A_\infty\}$
 - $L = \{\}$
- A_0 (Start) has no preconditions but all facts in the initial state as effects.
- A_∞ (Finish) has the goal conditions as preconditions and no effects.

Op(Action: Go(there); Precond: At(here);
Effects: At(there), \neg At(here))

Op(Action: Buy(x), Precond: At(store), Sells(store,x);
Effects: Have(x))



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POP Algorithm

- Stated as nondeterministic algorithm where choices must be made. Various search methods can be used (e.g. breadth-first, depth-first iterative deepening etc.) to explore the space of possible choices.
- Maintains agenda of goals that need to be supported by links, where an agenda element is a pair $\langle Q, A_i \rangle$ where Q is a precondition of A_i that needs supporting.
- Initialize plan to null plan and agenda to conjunction of goals (preconditions of Finish).
- Done when all preconditions of every action in plan are supported by causal links which are not threatened.

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POP($\langle A, O, L \rangle$, agenda)

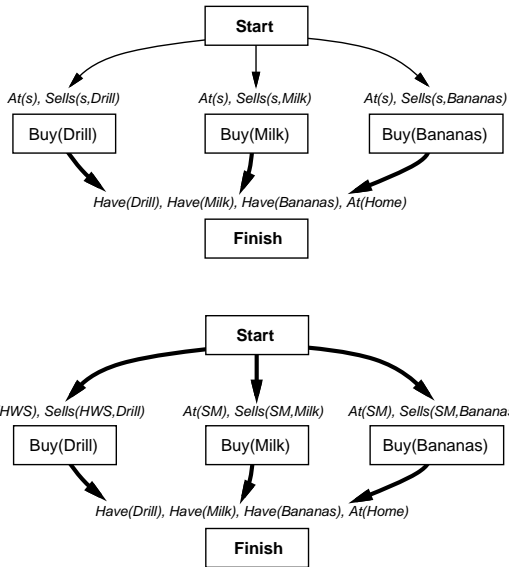
- 1) **Termination:** If agenda is empty, return $\langle A, O, L \rangle$. Use topological sort to determine a totally ordered plan.
- 2) **Goal Selection:** Let $\langle Q, A_{\text{need}} \rangle$ be a pair on the agenda
- 3) **Action Selection:** Let A_{add} be a nondeterministically chosen action that adds Q . It can be an existing action in A or a new action. If there is no such action return failure.

$$L' = L \cup \{A_{\text{add}} \rightarrow^Q A_{\text{need}}\}$$

$$O' = O \cup \{A_{\text{add}} < A_{\text{need}}\}$$
 if A_{add} is new
 then $A' = A \cup \{A_{\text{add}}\}$ and $O' = O' \cup \{A_0 < A_{\text{add}} < A_\infty\}$
 else $A' = A$
- 4) **Update goal set:** Let agenda' = agenda - $\{\langle Q, A_{\text{need}} \rangle\}$
 If A_{add} is new then for each conjunct Q_i of its precondition, add $\langle Q_i, A_{\text{add}} \rangle$ to agenda'
- 5) **Causal link protection:** For every action A_t that threatens a causal link $A_p \rightarrow^Q A_c$ add an ordering constraint by choosing nondeterministically either
 - (a) **Demotion:** Add $A_t < A_p$ to O'
 - (b) **Promotion:** Add $A_c < A_t$ to O'
 If neither constraint is consistent then return failure.
- 6) **Recurse:** POP($\langle A', O', L' \rangle$, agenda')

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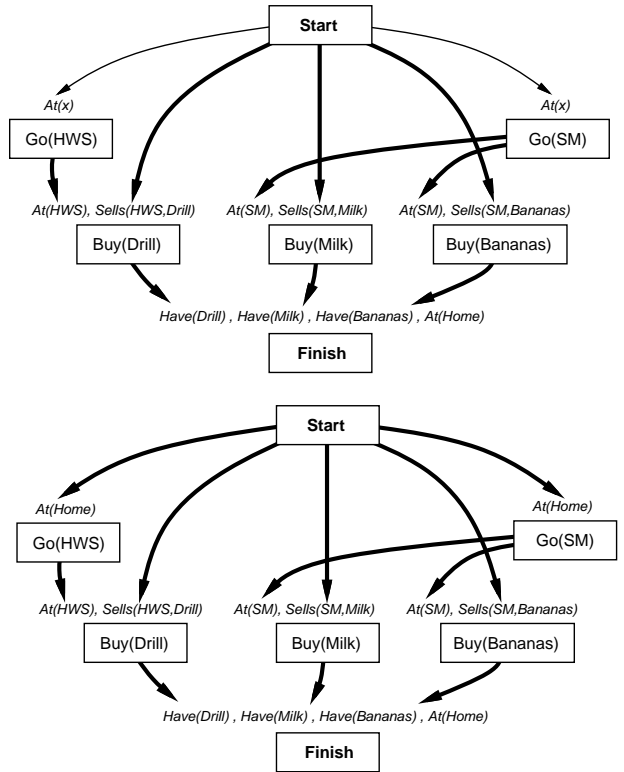
Example



- Add three actions to achieve basic goals. Use initial state to achieve the Sells preconditions.
- Bold links are causal, regular are just ordering constraints

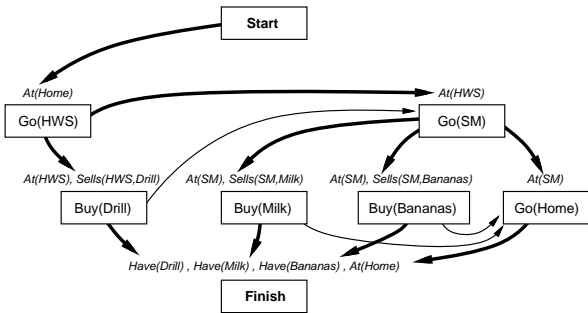
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Example (cont)



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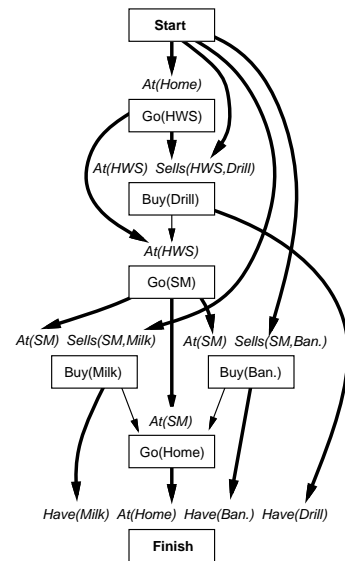
Example (cont)



- Cannot resolve threat to $At(Home)$ preconditions of both $Go(HWS)$ and $Go(SM)$.
- Must backtrack to supporting $At(x)$ precondition of $Go(SM)$ from initial state $At(Home)$ and support it instead from the $At(HWS)$ effect of $Go(HWS)$.
- Since $Go(SM)$ still threatens $At(HWS)$ of $Buy(Drill)$ must promote $Go(SM)$ to come after $Buy(Drill)$. Demotion is not possible due to causal link supporting $At(HWS)$ precondition of $Go(SM)$

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Example (cont)



- Add $Go(Home)$ action to achieve $At(Home)$, use $At(SM)$ to achieve its precondition, and order it after $Buy(Milk)$ and $Buy(Banana)$ to resolve threats to $At(SM)$.

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Planning Conclusions

- Experiments confirm that in most cases partial-order planning is more efficient than total order.
- Planning techniques have been applied to a number of realistic tasks:
 - Logistics planning for Desert Storm
 - Scheduling for the Hubble Space Telescope
 - Planning ground operations for the Space Shuttle
 - Semiconductor manufacturing
 - Cleaning oil spills
- Many issues are involved in interleaving planning and execution
 - Conditional planning
 - Execution monitoring and dynamic replanning