

Planning



- Search vs. planning
- STRIPS operators
- Partial-order planning

What we have so far



- Can TELL KB about new percepts about the world
- KB maintains model of the current world state
- Can ASK KB about any fact that can be inferred from KB

How can we use these components to build a [planning agent](#),

i.e., an agent that constructs plans that can achieve its goals, and that then executes these plans?

Example: Robot Manipulators

- Example: (courtesy of Martin Rohrmeier)
 - Puma 560
 - Kr6



Remember: Problem-Solving Agent

```
function SIMPLE-PROBLEM-SOLVING-AGENT(p) returns an action
  inputs: p, a percept
  static: s, an action sequence, initially empty
           state, some description of the current world state
           g, a goal, initially null
           problem, a problem formulation

  state ← UPDATE-STATE(state, p)
  if s is empty then
    g ← FORMULATE-GOAL(state)
    problem ← FORMULATE-PROBLEM(state, g)
    s ← SEARCH(problem)
  action ← RECOMMENDATION(s, state)
  s ← REMAINDER(s, state)
  return action
```

Note: This is *offline* problem-solving. *Online* problem-solving involves acting w/o complete knowledge of the problem and environment

Simple planning agent



- Use percepts to build model of current world state
- IDEAL-PLANNER: Given a goal, algorithm generates plan of action
- STATE-DESCRIPTION: given percept, return initial state description in format required by planner
- MAKE-GOAL-QUERY: used to ask KB what next goal should be

A Simple Planning Agent

```
function SIMPLE-PLANNING-AGENT(percept) returns an action
  static:
    KB, a knowledge base (includes action descriptions)
    p, a plan (initially, NoPlan)
    t, a time counter (initially 0)

  local variables:G, a goal
    current, a current state description

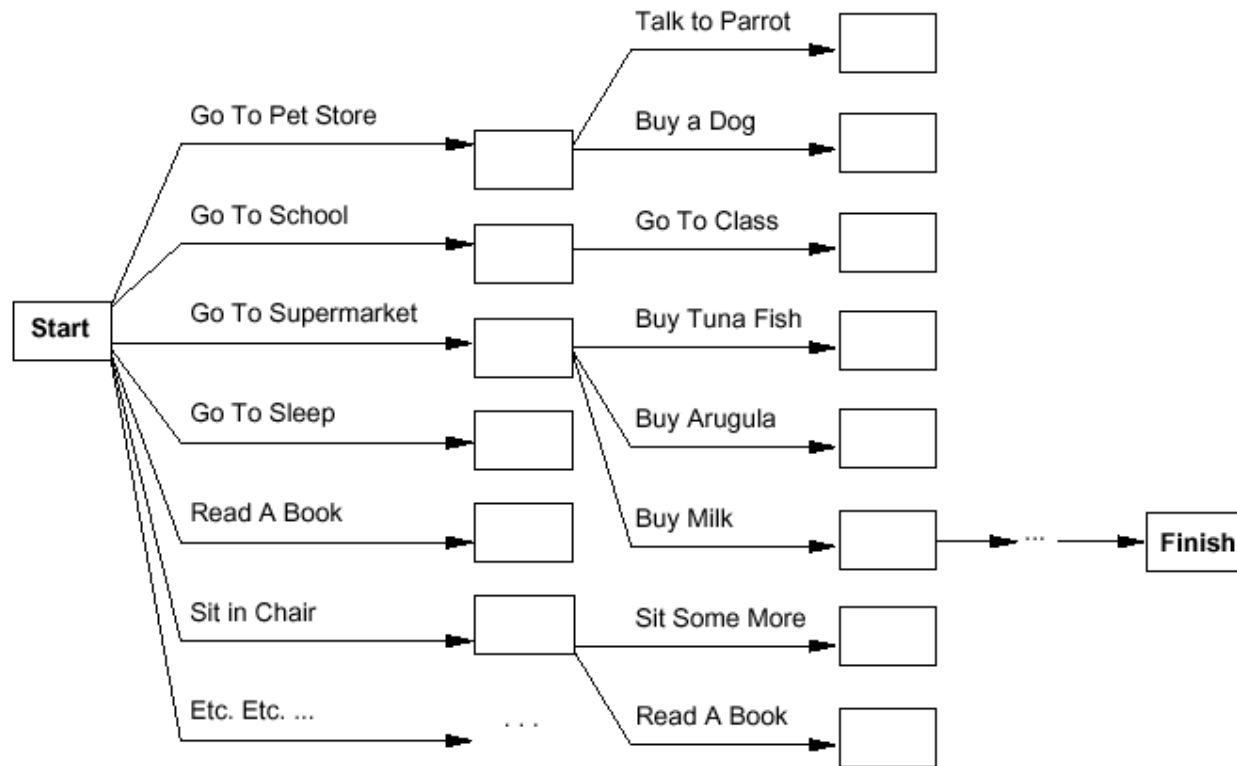
  TELL(KB, MAKE-PERCEPT-SENTENCE(percept, t))
  current ← STATE-DESCRIPTION(KB, t)
  if p = NoPlan then
    G ← ASK(KB, MAKE-GOAL-QUERY(t))
    p ← IDEAL-PLANNER(current, G, KB)
  if p = NoPlan or p is empty then
    action ← NoOp
  else
    action ← FIRST(p)
    p ← REST(p)
  TELL(KB, MAKE-ACTION-SENTENCE(action, t))
  t ← t+1
  return action
```

Like popping from a stack

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

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	Lisp data structures	Logical sentences
Actions	Lisp code	Preconditions/outcomes
Goal	Lisp 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

Basic representation for planning

- Most widely used approach: uses STRIPS language
- **states**: conjunctions of function-free ground literals (I.e., predicates applied to constant symbols, possibly negated); e.g.,

$\text{At}(\text{Home}) \wedge \neg\text{Have}(\text{Milk}) \wedge \neg\text{Have}(\text{Bananas}) \wedge \neg\text{Have}(\text{Drill}) \dots$

- **goals**: also conjunctions of literals; e.g.,

$\text{At}(\text{Home}) \wedge \text{Have}(\text{Milk}) \wedge \text{Have}(\text{Bananas}) \wedge \text{Have}(\text{Drill})$

but can also contain variables (implicitly universally quant.); e.g.,

$\text{At}(x) \wedge \text{Sells}(x, \text{Milk})$

Planner vs. theorem prover



- **Planner:** ask for sequence of actions that makes goal true if executed
- **Theorem prover:** ask whether query sentence is true given KB

STRIPS operators

Tidily arranged actions descriptions, restricted language

ACTION: $Buy(x)$

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

EFFECT: $Have(x)$

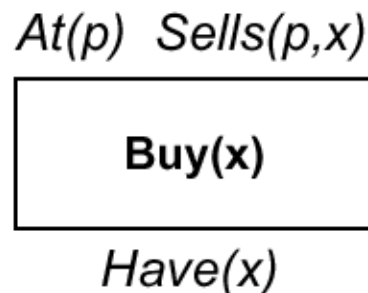
[Note: this abstracts away many important details!]

Restricted language \Rightarrow efficient algorithm

Precondition: conjunction of positive literals

Effect: conjunction of literals

Graphical notation:



Types of planners



- Situation space planner: search through possible situations
- Progression planner: start with initial state, apply operators until goal is reached
 - Problem: high branching factor!
- Regression planner: start from goal state and apply operators until start state reached
 - Why desirable? usually many more operators are applicable to initial state than to goal state.
 - Difficulty: when want to achieve a conjunction of goals

Initial STRIPS algorithm: situation-space regression planner

State space vs. plan space

Standard search: node = concrete world state

Planning search: node = partial plan Search space of plans rather than of states.

Defn: open condition is a precondition of a step not yet fulfilled

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

Operations on plans



- Refinement operators: add constraints to partial plan

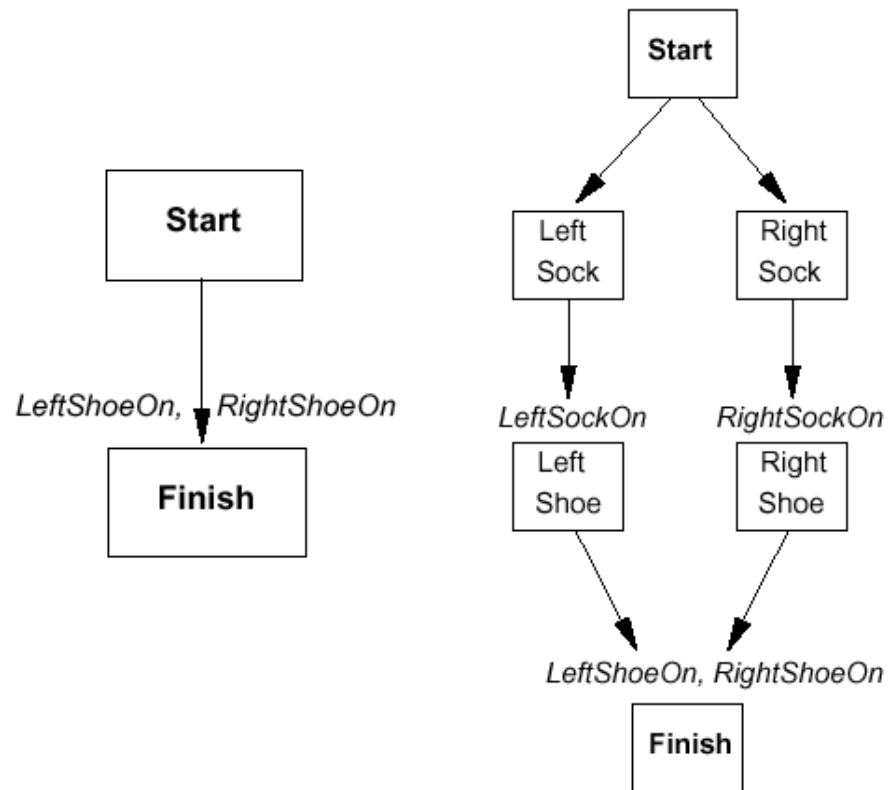
- Modification operator: every other operators

Types of planners



- **Partial order planner:** some steps are ordered, some are not
- **Total order planner:** all steps ordered (thus, plan is a simple list of steps)
- **Linearization:** process of deriving a totally ordered plan from a partially ordered plan.

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

Plan

We formally define a plan as a **data structure consisting of**:

- Set of **plan steps** (each is an operator for the problem)
- Set of **step ordering constraints**

e.g., $A < B$ means "A before B"

- Set of **variable binding constraints**

e.g., $v = x$ where v variable and x constant or other variable

- Set of **causal links**

e.g., $A \xrightarrow{c} B$ means "A achieves c for B"

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 ← SELECT-SUBGOAL(*plan*)

 CHOOSE-OPERATOR(*plan, operators, S_{need}, 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 (cont.)

procedure CHOOSE-OPERATOR($plan, operators, S_{need}, c$)

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**

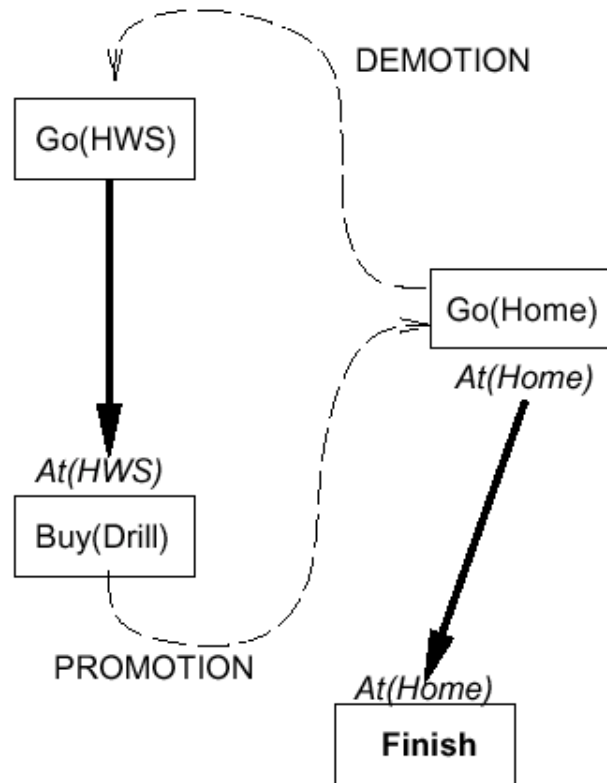
end

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

Extensions for disjunction, universals, negation, conditionals

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 $At(HWS)$:



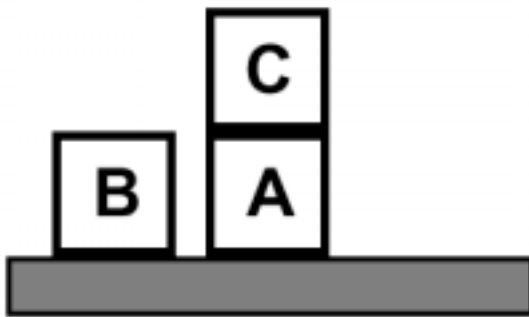
Demotion: put before $Go(HWS)$

Promotion: put after $Buy(Drill)$

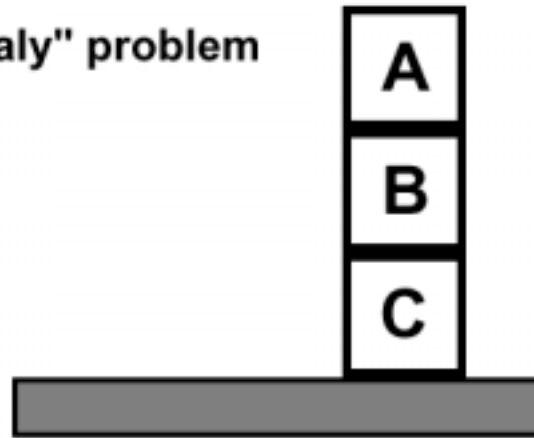
Example: block world



"Sussman anomaly" problem

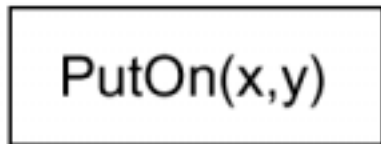


Start State



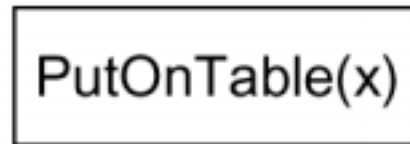
Goal State

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



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

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



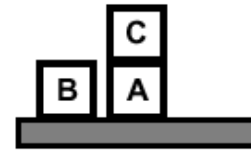
$\sim On(x,z) \ Clear(z) \ On(x, Table)$

+ several inequality constraints

Example (cont.)

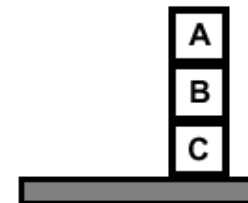
START

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

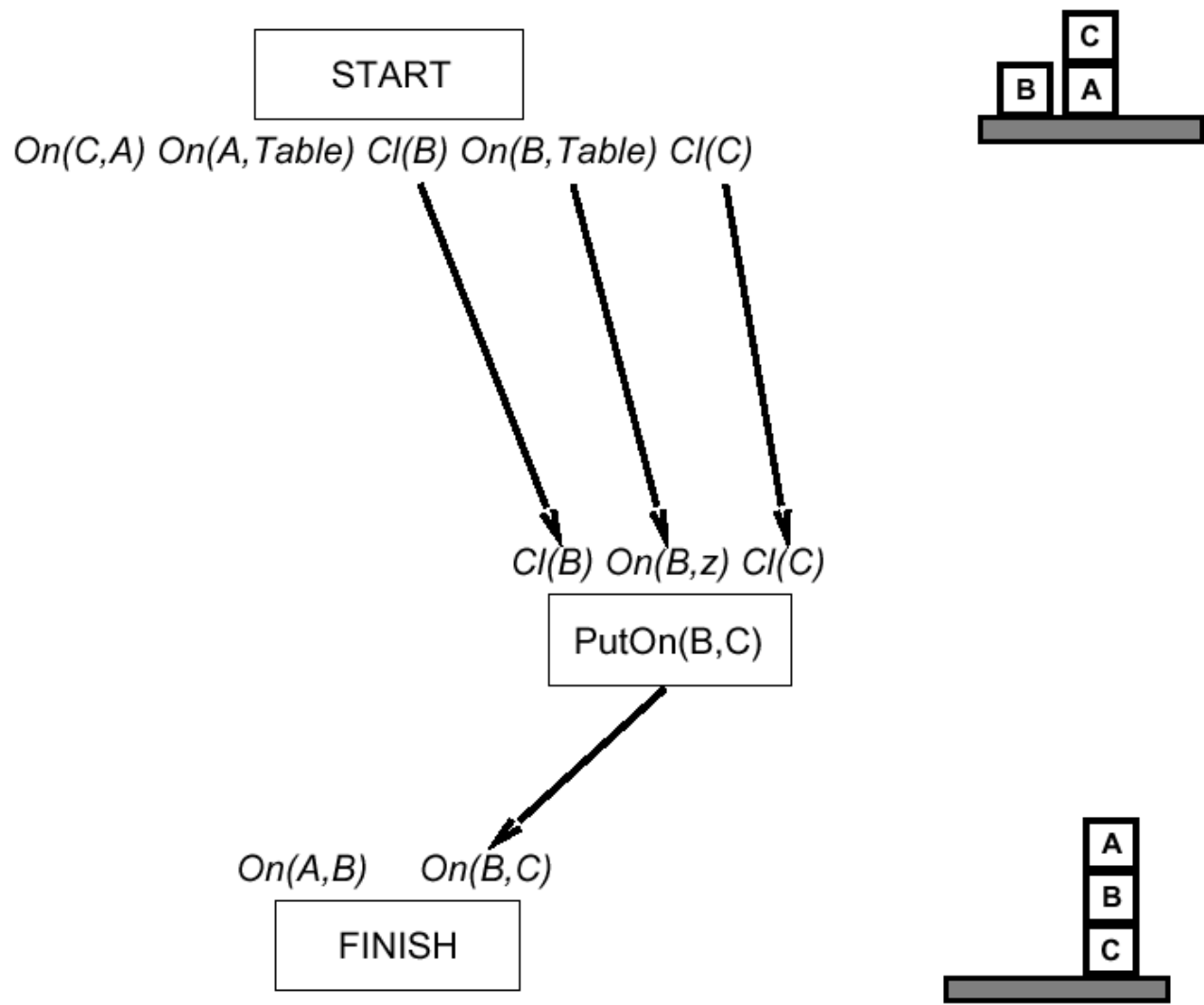


$On(A,B)$ $On(B,C)$

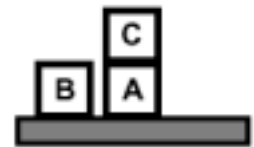
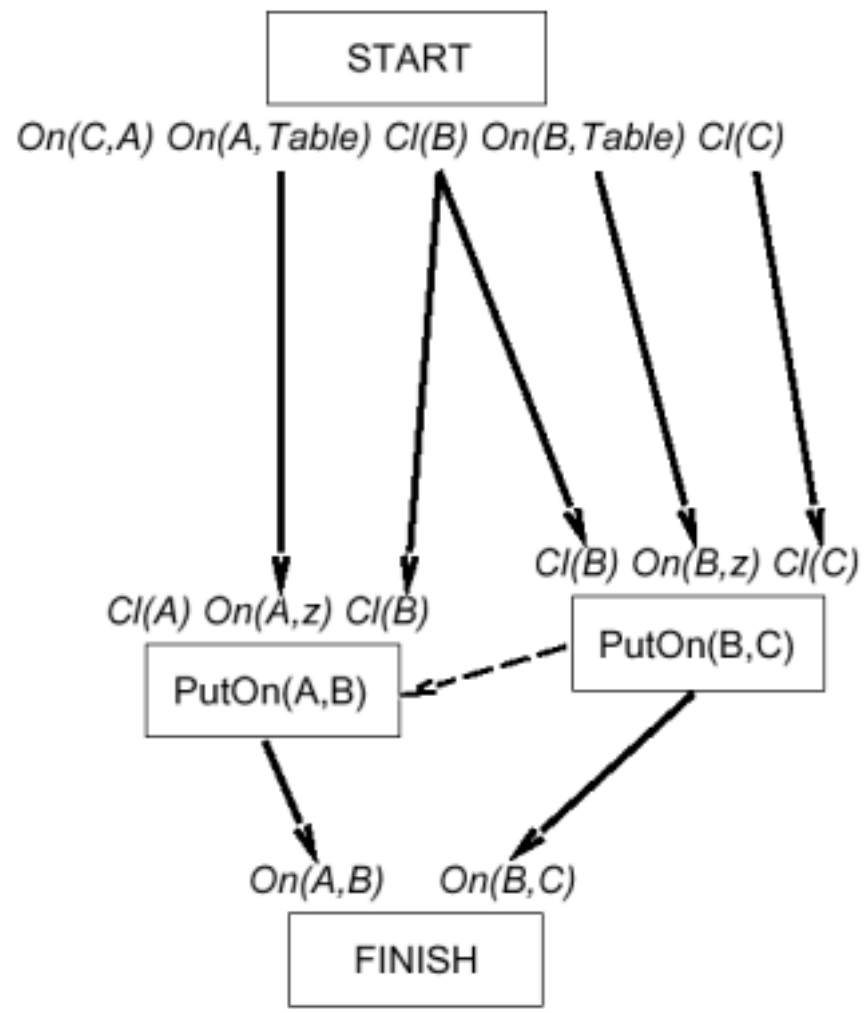
FINISH



Example (cont.)



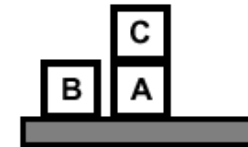
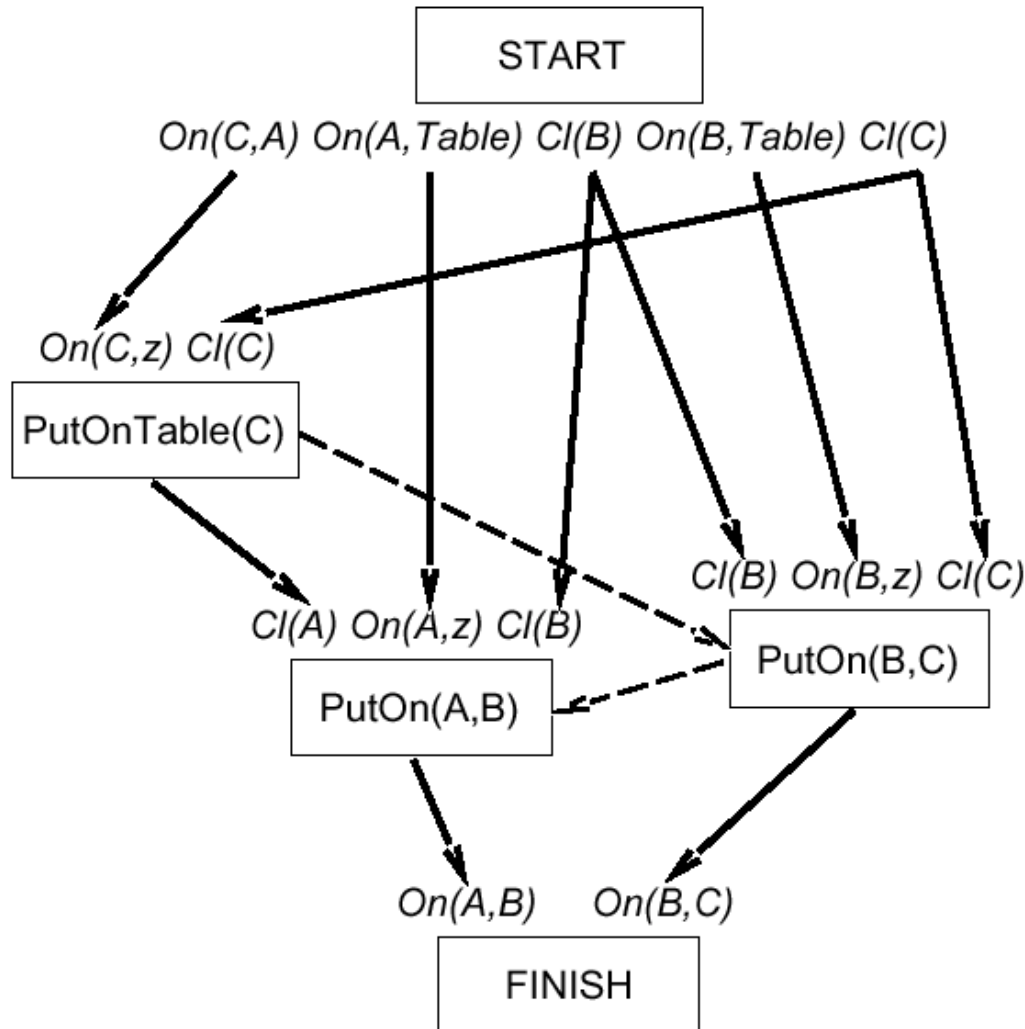
Example (cont.)



PutOn(A,B)
 clobbers Cl(B)
 => order after
 PutOn(B,C)



Example (cont.)



PutOn(A,B)
 clobbers Cl(B)
 => order after
 PutOn(B,C)

PutOn(B,C)
 clobbers Cl(C)
 => order after
 PutOnTable(C)

