

Planning

Chapter 11

AI – A Modern Approach.

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Review

- ❑ What is Planning?
- ❑ STRIPS Language and Representing planning problems using STRIPS.
- ❑ Progression and regression algorithm.
- ❑ Heuristics for progression and regression algorithm
- ❑ Total ordered planners and partial order planners
- ❑ Keywords in partial order planners.

POP Algorithm

- Starts with the definition of initial state, actions, goal test.
- Initial plan contains Start and Finish actions, the ordering constraint $Start \prec Finish$ and no causal links and has all the preconditions in Finish as open preconditions.

POP Algorithm Contd.

- Successor Function: Arbitrarily picks an open precondition p on an action B and generates successor plan.
- Consistency enforcement:
 - The causal link $A \xrightarrow{p} B$ and the ordering constraint $A < B$ are added to the plan.
 - Resolve conflicts between the new causal link and all existing actions. E.g. if action C conflicts with $A \xrightarrow{p} B$ then it is resolved by making C to occur at sometime before action A ($C < A$) or after action B ($B < C$) by adding ordering constraints. Add all successor states for either or both if they result in consistent plans.

POP Algorithm Contd.

- Backtrack if an open condition is unachievable or if a conflict is irresolvable.
- Goal test: Checks whether a plan is a solution to the original planning problem by checking if there any open preconditions left. If set of open preconditions set is empty then POP has reached a solution.

POP Example

Agent is at hardware store (HWS). Agent has to buy milk and eggs from supermarket and return home.

Start State:

At (HWS) \wedge \neg Have (Milk) \wedge \neg Have (Eggs) \wedge
Sells (SM, Milk) \wedge Sells (SM, Eggs)

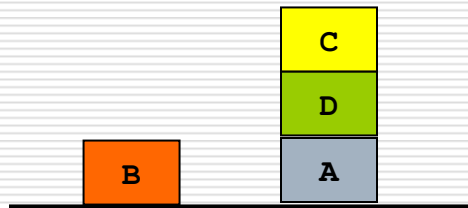
Goal State:

At (Home) \wedge Have (Milk) \wedge Have (Eggs)

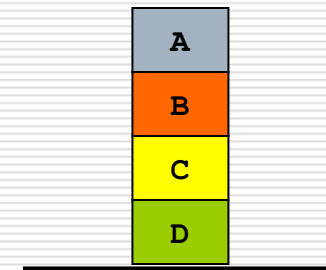
[\(Example using POP\)](#)

Blocks World Example using POP

Initial State



Goal State



Action :**Move** (p, x, y)

Precond: $\text{On}(p, x) \wedge \text{clear}(y) \wedge \text{clear}(p)$

Effect : $\text{On}(p, y) \wedge \text{clear}(x) \wedge \neg \text{clear}(y) \wedge \neg \text{On}(p, x)$

Op(Action :**Movetotable** (p, x)

Precond: $\text{On}(p, x) \wedge \text{clear}(p)$

Effect : $\text{On}(p, \text{Table}) \wedge \text{clear}(x) \wedge \neg \text{On}(p, x)$

[\(Blocks World Example using POP\)](#)

Heuristics for POP

- Count the number of distinct open preconditions.
 - Overestimates: When start state has literal that matches open precondition in finish state.
 - Underestimates: when there is negative interactions between actions.

Heuristics for POP

- Better approach to calculate heuristics:
 - Choose open preconditions that can be satisfied in the fewest number of ways.

Planning Graphs

- Used to achieve better heuristic estimates.
 - A solution can also directly extracted using GRAPHPLAN.
- Consists of a sequence of levels that correspond to time steps in the plan.
 - Level 0 is the initial state.
 - Each level consists of a set of literals and a set of actions.
 - *Literals* = all those that **could be true** at that time step, depending upon the actions executed at the preceding time step.
 - *Actions* = all those actions that **could have their preconditions satisfied** at that time step, depending on which of the literals actually hold.

Cake Example

Init (Have (Cake))

Goal (Have (Cake) \wedge Eaten (Cake))

Action (**Eat (Cake)**)

PRECOND: Have (Cake)

EFFECT:

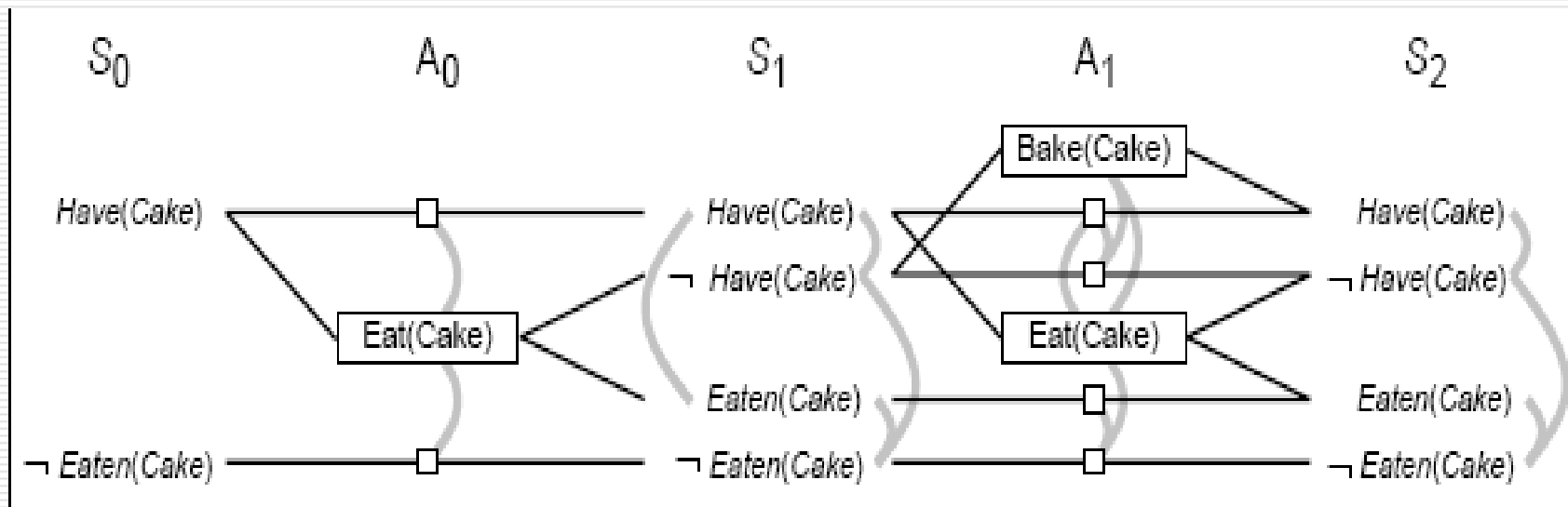
\neg Have (Cake) \wedge Eaten (Cake))

Action (**Bake (Cake)**)

PRECOND: Have (Cake)

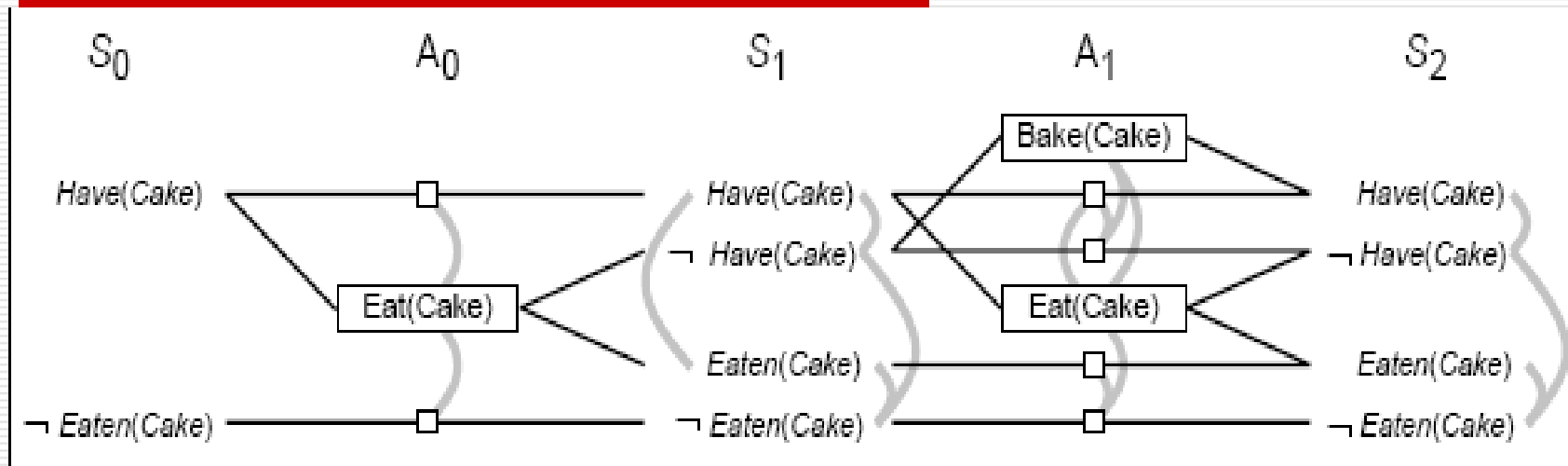
EFFECT: Have (Cake))

Cake Example



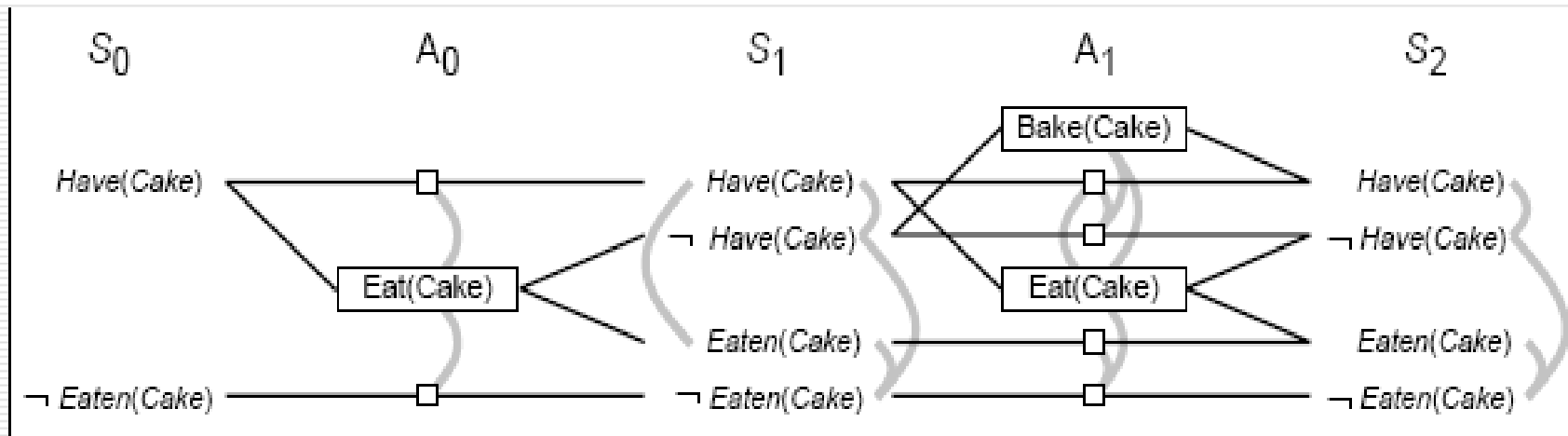
- Start at level S_0 and determine action level A_0 and next level S_1 .
 - A_0 contains all actions whose preconditions are satisfied in the previous level.

Cake Example



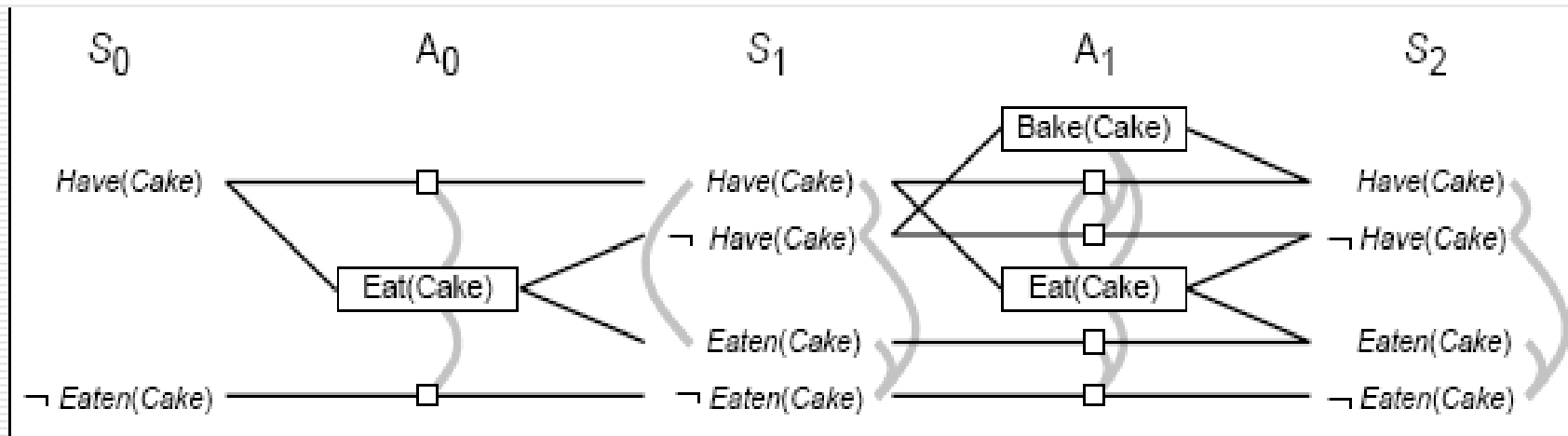
- Connect precondition and effect of actions.
- Inaction is represented by persistence actions.
- Level A0 contains the actions that could occur
 - Conflicts between actions are represented by *mutex* links

Cake Example



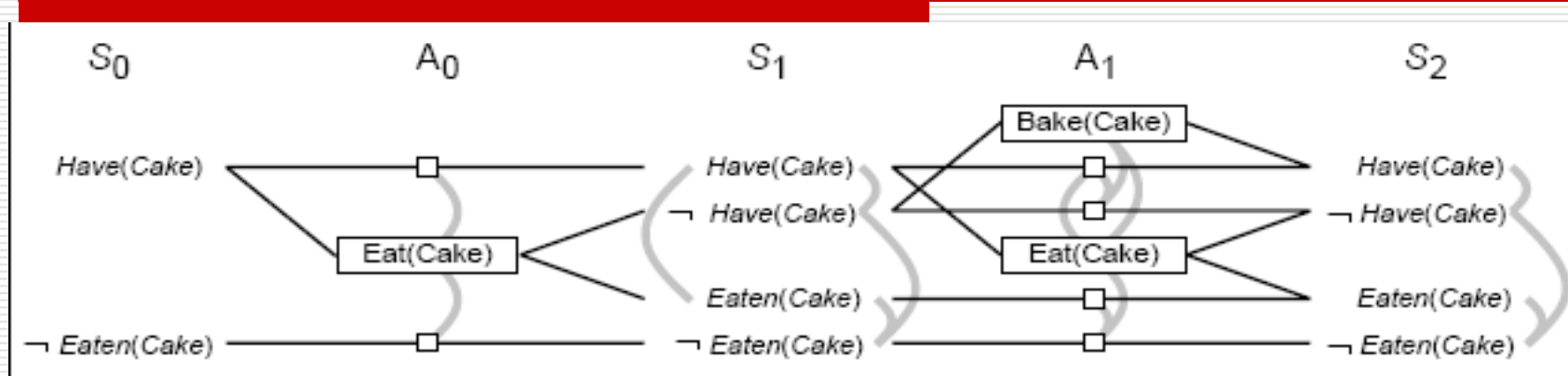
- Level S_1 contains all literals that could result from picking any subset of actions in A_0
 - Conflicts between literals that can not occur together (as a consequence of the selection action) are represented by mutex links.

Cake Example



- S1 defines multiple states and the mutex links are the constraints that define this set of states.
- Continue until two consecutive levels are identical

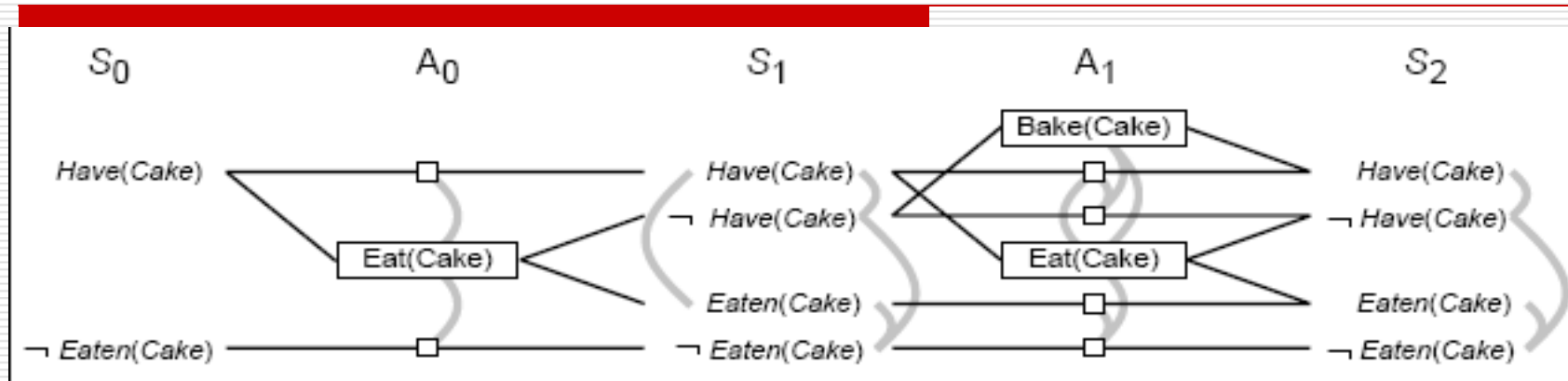
Cake Example



□ A mutex relation holds between two actions when:

- 1. Inconsistent effects:** one action negates the effect of another.
- 2. Interference:** one of the effects of one action is the negation of a precondition of the other.

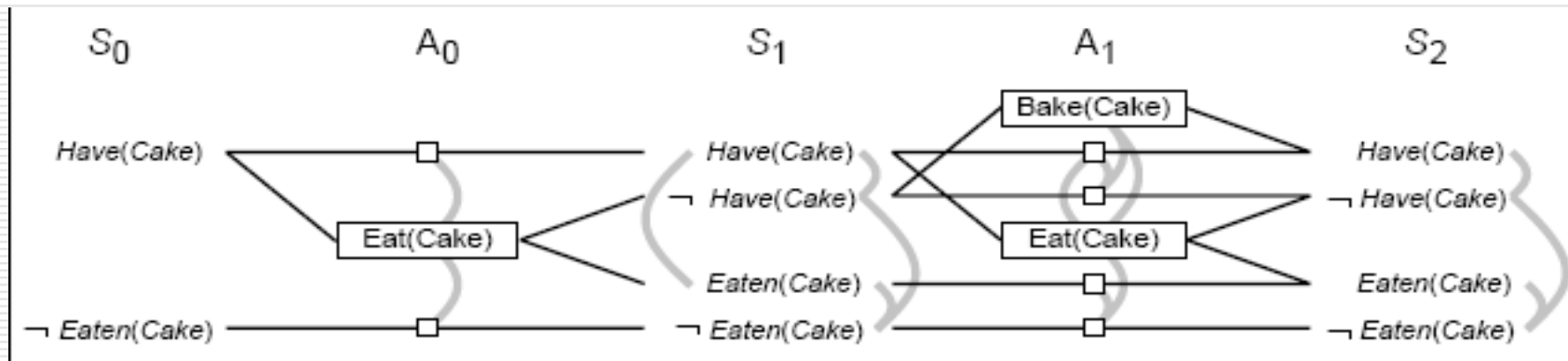
Cake Example



□ A mutex relation holds between two actions when:

3. Competing needs: one of the preconditions of one action is mutually exclusive with the precondition of the other.

Cake Example



- A mutex relation holds between two literals when **(inconsistent support)**:
1. If one is the negation of the other
 2. If each possible action pair that could achieve the literals is mutex.

Planning Graphs for Heuristic Estimation

- A literal that does not appear in the final level of the graph cannot be achieved by any plan.
 - Useful for backward search (cost = inf).
- Level of appearance can be used as cost estimate of achieving any goal literals = *level cost*.
- Small problem: several actions can occur
 - Restrict to one action using serial PG (add mutex links between every pair of actions, except persistence actions).
- Cost of a conjunction of goals? Max-level, sum-level and set-level heuristics.

GraphPlan Algorithm

- extracts a solution directly from the PG

```
function GRAPHPLAN(problem) returns solution or failure
  graph ← INITIAL-PLANNING-GRAPH(problem)
  goals ← GOALS[problem]
  loop do
    if goals all non-mutex in last level of graph then do
      solution ← EXTRACT-SOLUTION(graph, goals, LENGTH(graph))
      if solution ≠ failure then return solution
      else if NO-SOLUTION-POSSIBLE(graph) then return failure
    graph ← EXPAND-GRAPH(graph, problem)
```

- **EXTRACT-SOLUTION:**

- checks whether a plan can be found searching backwards

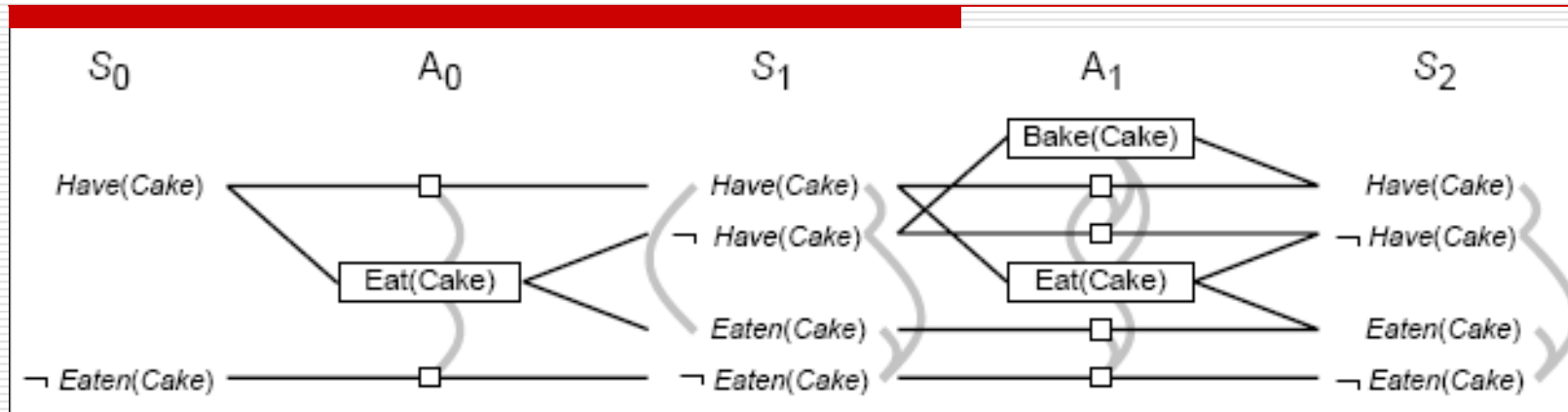
- **EXPAND-GRAPH:**

- adds actions for the current and state literals for the next level

Extract Solution

- **A state consists of**
 - a pointer to a level in the planning graph
 - a set of unsatisfied goals
- **Initial state**
 - last level of PG
 - set of goals from the planning problem
- **Actions**
 - select any set of non-conflicting subset of the actions of A_{i-1} that cover the goals in the state
- **Goal**
 - success if level S_0 is reached with such with all goals satisfied

Cake example



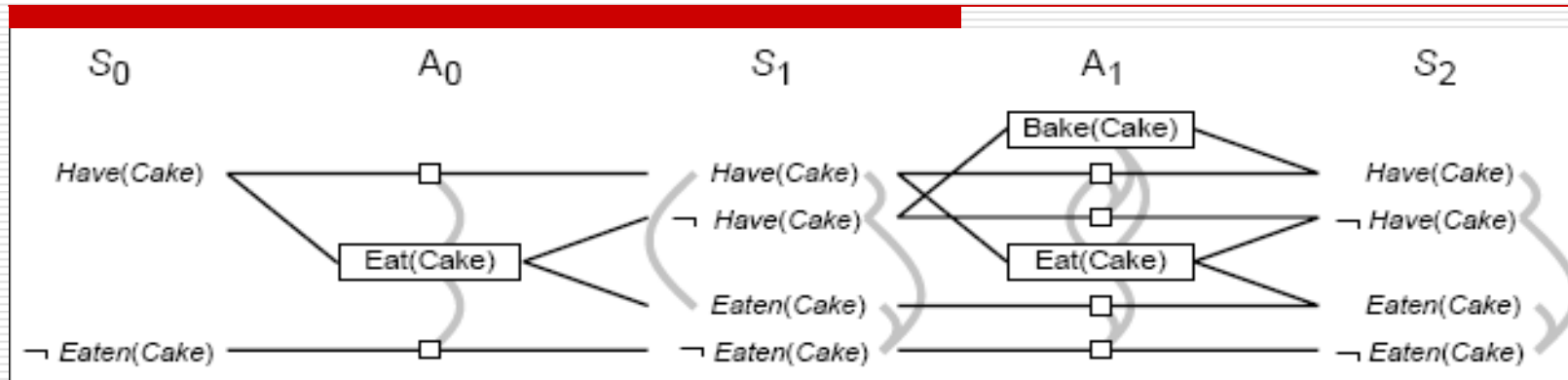
- Start with Goal state (literals):

$Have(Cake) \wedge Eaten(Cake)$ in S_2

- Only non conflicting Action choices are: Bake (Cake) , Persistent action (Eaten (Cake)).

as all the other have mutex relation with respect to either their preconditions or effects.

Cake example



□ Literals at S_1 :

$\neg Have(Cake) \wedge \neg Eaten(Cake)$.

□ Only action:

$Eat(Cake)$

□ Literals at S_0 :

$Have(Cake) \wedge \neg Eaten(Cake)$. (graphplan terminates)

Planning with Propositional Logic

- Planning can be done by proving theorem in situation calculus.
- Here: test the *satisfiability* of a logical sentence:

initial state \wedge *all possible action descriptions* \wedge *goal*

- Sentence contains propositions for every action occurrence.
 - A model will assign true to the actions that are part of the correct plan and false to the others
 - An assignment that corresponds to an incorrect plan will not be a model because of inconsistency with the assertion that the goal is true.
 - If the planning is unsolvable the sentence will be unsatisfiable.