

# Planning

- Introduction
  - Planning vs. Problem-Solving
  - Representation in Planning Systems
- Situation Calculus
  - The Frame Problem
- STRIPS representation language
  - Blocks World
- Planning with State-Space Search
  - Progression Algorithms
  - Regression Algorithms
- Planning with Plan-Space Search
  - Partial-Order Planning
  - The Plan Graph and GraphPlan
  - SatPlan

Material from  
Russell & Norvig,  
chapters 10.3. and 11

Slides based on Slides  
by Russell/Norvig,  
Lise Getoor  
and Tom Lenaerts

# Partial-Order Planning (POP)

- Progression and regression planning are **totally ordered** plan search forms
  - this means that in all searched plans the sequence of actions is completely ordered
  - Decisions must be made on how to sequence actions in all the subproblems
    - They cannot take advantage of problem decomposition
- If actions do not interfere with each other, they could be made in any order (or in parallel) → **partially ordered** plan
  - if a plan for each subgoal only makes minimal commitments to orders
    - only orders those actions that must be ordered for a successful completion of the plan
  - it can re-order steps later on (when subplans are combined)
  - **Least commitment strategy:**
    - Delay choice during search

# Shoe Example

Initial State: nil  
Goal State: RightShoe & LeftShoe

```
Action( LeftSock,  
PRECOND: -  
ADD:     LeftSockOn  
DELETE:  -  
)
```

```
Action( RightSock,  
PRECOND: -  
ADD:     RightSockOn  
DELETE:  -  
)
```

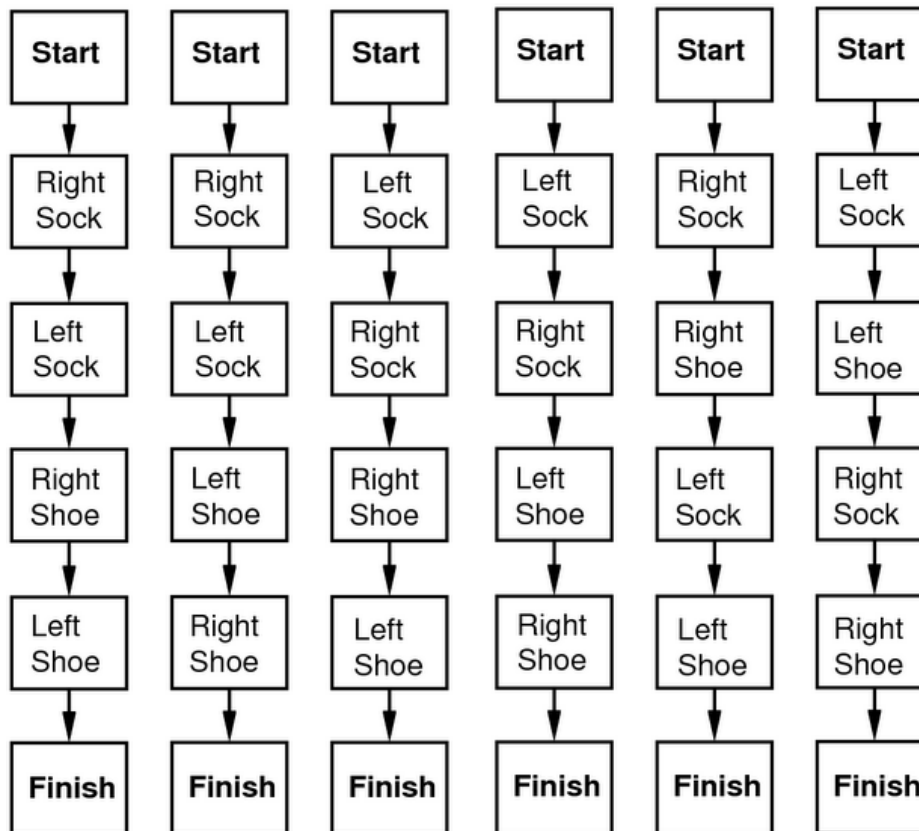
```
Action( LeftShoe,  
PRECOND: LeftSockOn  
ADD:     LeftShoeOn  
DELETE:  -  
)
```

```
Action( RightShoe,  
PRECOND: RightSockOn  
ADD:     RightShoeOn  
DELETE:  -  
)
```

# Shoe Example

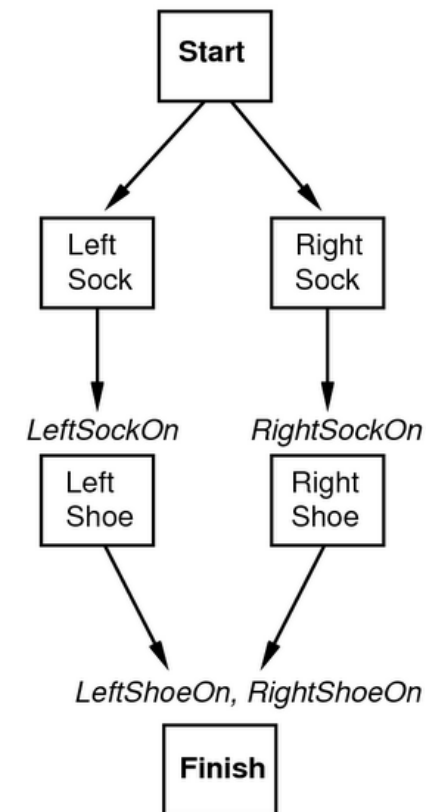
## ■ Total-Order Planner

- all actions are completely ordered



## ■ Partial-Order Planner

- may leave the order of some actions undetermined
- any order is valid



# POP as a Search Problem

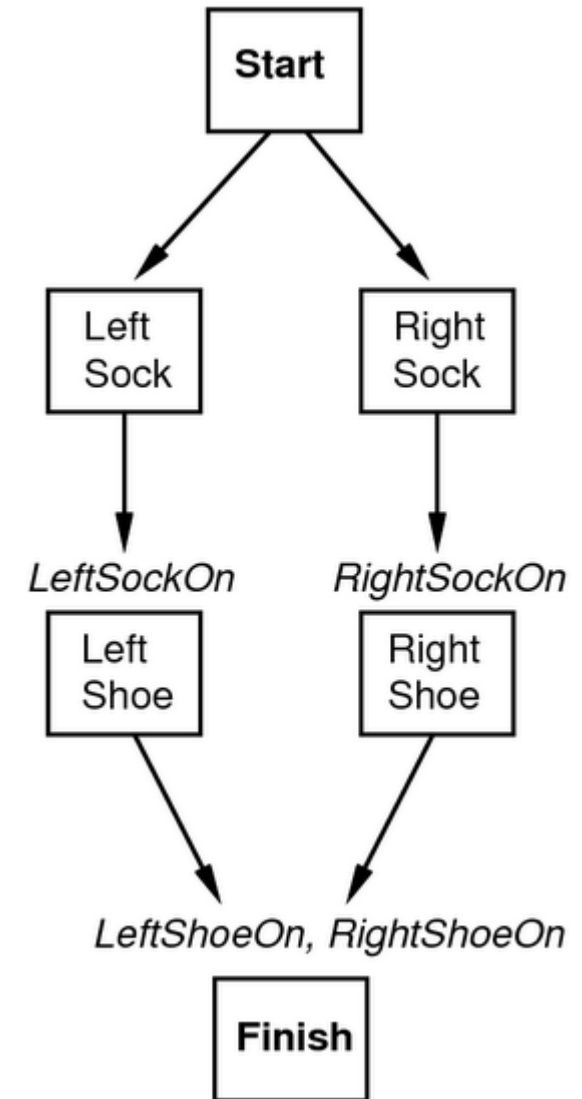
- A solution can be found by a **search through Plan-Space**:
  - **States are** (mostly unfinished) **plans**

Each plan has 4 components:

- A set of **actions** (steps of the plan)
- A set of **ordering constraints**:  $A < B$  ( $A$  before  $B$ )
  - Cycles represent contradictions.
- A set of **causal links**  $A \rightarrow p \rightarrow B$  ( $A$  adds  $p$  for  $B$ )
  - The plan may not be extended by adding a new action  $C$  that conflicts with the causal link.
    - if the effect of  $C$  is  $\neg p$  and if  $C$  could come after  $A$  and before  $B$
- A set of **open preconditions**
  - Preconditions that are not achieved by action in the plan

# Example of Final Plan

- Actions = {RightSock, RightShoe, LeftSock, LeftShoe, Start, Finish}
- Orderings =
   
 { RightSock < RightShoe;
   
 LeftSock < LeftShoe}
- Links =
   
 { RightSock → RightSockOn → RightShoe,
   
 LeftSock → LeftSockOn → LeftShoe,
   
 RightShoe → RightShoeOn → Finish,
   
 ...}
- Open preconditions = {}



# Search through Plan-Space

- **Initial State** (empty plan):
  - contains only virtual *Start* and *Finish* actions
  - ordering constraint  $Start < Finish$
  - no causal links
  - all preconditions in *Finish* are open (these are the original goal)
- **Successor Function** (refining the plan):
  - picks one open precondition  $p$  on an action  $B$
  - generates a successor plan for every possible *consistent* way of choosing action that achieves  $p$
  - a plan is **consistent** iff there are no cycles in the ordering constraints and no conflicts with the causal links
- **Goal test:**
  - A consistent plan with no open preconditions is a solution.

# Subroutines

- **Refining a plan** with action  $A$ , which achieves  $p$  for  $B$ :
  - add causal link  $A \rightarrow p \rightarrow B$
  - add the ordering constraint  $A < B$
  - add  $Start < A$  and  $A < Finish$  to the plan (only if  $A$  is new)
  - resolve conflicts between
    - new causal link and all existing actions
    - between action  $A$  and all existing causal links (only if  $A$  is new)
- **Resolving a conflict** between a **causal link**  $A \rightarrow p \rightarrow B$  and an **action**  $C$ 
  - we have a conflict if the effect of  $C$  is  $\neg p$  and  $C$  could come after  $A$  and before  $B$
  - resolved by adding the ordering constraints  $C < A$  or  $B < C$ 
    - both refinements are added (two successor plans) if both are consistent



# Search through Plan-Space

- **Operators** on partial plans
  - Add an action to fulfill an open condition
  - Add a causal link
  - Order one step w.r.t another to remove possible conflicts
- **Search** gradually moves from incomplete/vague plans to complete/correct plans
- **Backtrack** if an open condition is unachievable or if a conflict is irresolvable
  - pick the next condition to achieve at one of the previous choice points
  - ordering of the conditions is irrelevant for completeness (the same plans will be found), but may be relevant for consistency

# Executing Partially Ordered Plans

- Any particular order that is consistent with the ordering constraints is possible
  - A partial order plan is executed by repeatedly choosing any of the possible next actions.
- This flexibility is a benefit in non-cooperative environments.

# Example: Spare Tire Problem

Initial State:    `at(flat, axle) ,`  
                   `at(spare, trunk)`

Goal State:      `at(spare, axle)`

```
Action( remove(spare, trunk) ,
PRECOND: at(spare, trunk)
ADD:     at(spare, ground)
DELETE:  at(spare, trunk)
)
```

```
Action( leave-overnight,
PRECOND: -
ADD:     -
DELETE:  at(spare, ground) ,
          at(spare, axle) ,
          at(spare, trunk) ,
          at(flat, ground) ,
          at(flat, axle)
)
```

```
Action( remove(flat, axle) ,
PRECOND: at(flat, axle)
ADD:     at(flat, ground)
DELETE:  at(flat, axle)
)
```

```
Action( putOn(spare, axle) ,
PRECOND: at(spare, ground) ,
          not(at(flat, axle)) ,
          at(spare, axle)
ADD:     at(spare, axle)
DELETE:  at(spare, ground)
)
```

Here we need a `not`, which is not part of the original STRIPS language!

# Example: Spare Tire Problem

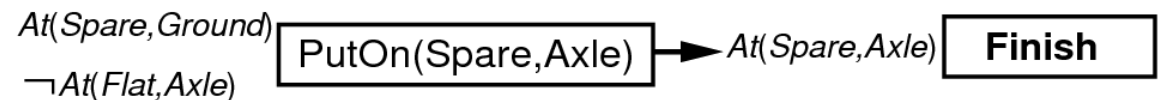
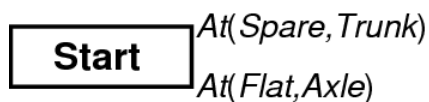
- Initial plan:
  - Action **start** has the current state as effects
  - Action **finish** has the goal as preconditions

**Start**  $At(Spare, Trunk)$   
 $At(Flat, Axle)$

$At(Spare, Axle)$  **Finish**

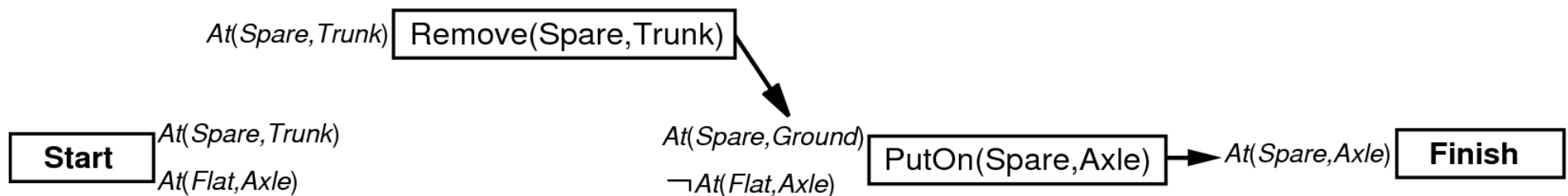
# Example: Spare Tire Problem

- Action `putOn (spare , axle)` is the only action that achieves the goal `at (spare , axle)`
- the current plan is refined to one new plan:
  - `putOn (spare , axle)` is added to the list of actions
  - add constraint `putOn (spare , axle) < finish`
  - add causal link `putOn (spare , trunk) → at (spare , axle) → finish`
  - the preconditions of `putOn (spare , trunk)` are now open



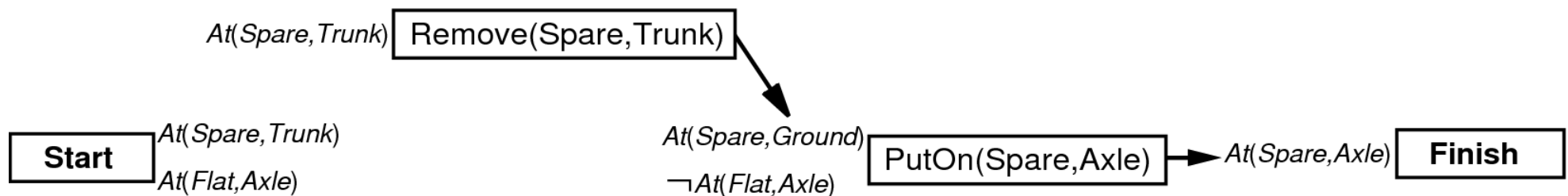
# Example: Spare Tire Problem

- we select the next open precondition **at (spare , ground)** as a goal
- only **at (spare , ground)** can achieve this goal
- the current plan is refined to a new one as before



# Example: Spare Tire Problem

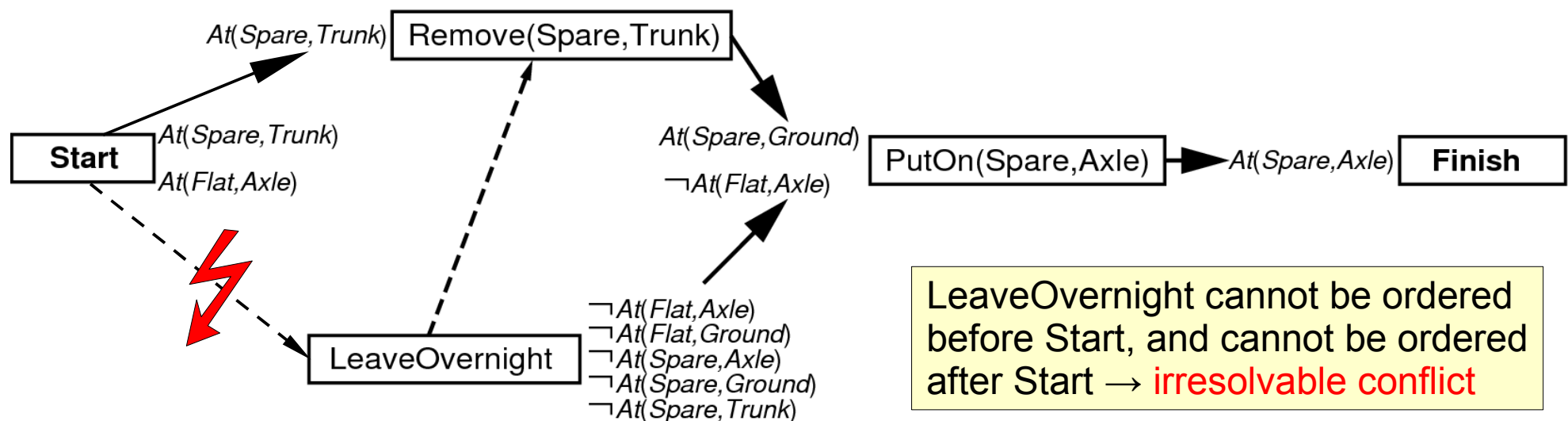
- we select the next open precondition `not (at (flat, axle))` as a goal
  - could be achieved with two actions
    - `leave-overnight`
    - `remove (flat, axle)`
- we have two successor plans



# Example: Spare Tire Problem

## Plan 1: **leave-overnight**

- is in conflict with the constraint  
 $\text{remove}(\text{spare}, \text{trunk}) \rightarrow \text{remove}(\text{spare}, \text{trunk}) \rightarrow \text{putOn}(\text{spare}, \text{axle})$   
 $\rightarrow$  has to be ordered before  $\text{remove}(\text{spare}, \text{trunk})$
- the condition  $\text{at}(\text{spare}, \text{trunk})$  has to be achieved next
  - start** is the only action that can achieve this
  - however, **start** is in conflict with **leave-overnight**
  - this conflict cannot be resolved  $\rightarrow$  **backtracking**

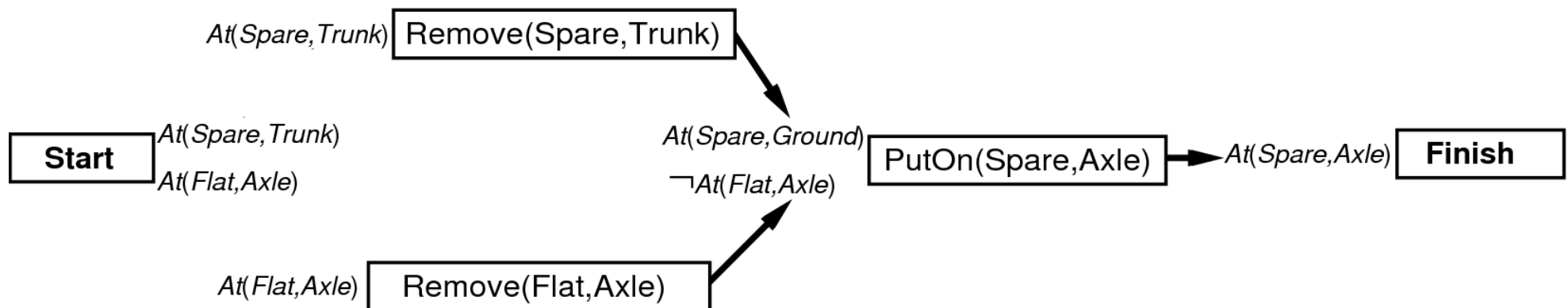




# Example: Spare Tire Problem

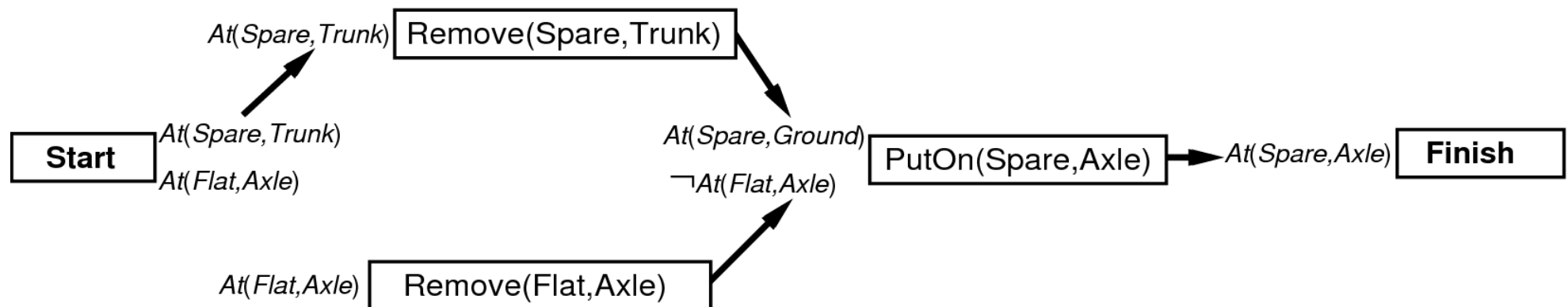
Plan 2: **remove (flat, axle)**

- achieves goal **not (at (flat, axle) )**
- corresponding causal link and order relation are added
- **at (flat, axle)** becomes open precondition



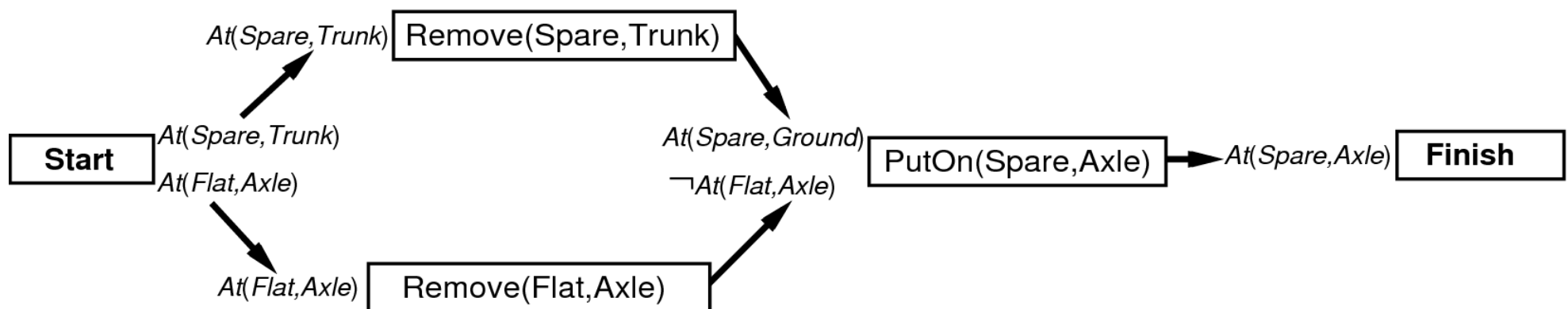
# Example: Spare Tire Problem

- open precondition **at (spare , trunk)** is selected as goal
  - action **start** is added
  - corresponding causal link and order relation are added



# Example: Spare Tire Problem

- open precondition **at (spare, trunk)** is selected as goal
  - action **start** is added
  - corresponding causal link and order relation are added
- open precondition **at (flat, axle)** is selected as goal
  - action **start** is added
  - corresponding causal link and order relation are added
- no more open preconditions remain
  - **plan is completed**



# POP in First-Order Logic

- Operators may leave some variables unbound
- Example**
  - Achieve goal `on(a, b)` with action `move(a, From, b)`
  - It remains unspecified from where block `a` should be moved (`PRECOND: on(a, From)`)

```

Action( move(Block, From, To) ,
PRECOND: on(Block, From) ,
        clear(Block) ,
        clear(To) ,
ADD:    on(Block, To) ,
        clear(From) ,
DELETE: on(Block, From) ,
        clear(To)
)

```

- Two approaches**
  - Decide for one binding and **backtrack** later on (if necessary)
  - Defer the choice for later (**least commitment**)
- Problems** with least commitment:
  - e.g., an action that has `on(a, From)` on its delete-list will only conflict with above if both are bound to the same variable
  - can be resolved by introducing inequality constraint.

# Heuristics for Plan-Space Planning

- Not as well understood as heuristics for state-space planning
- **General heuristic**: number of distinct open preconditions
  - maybe minus those that match the initial state
  - underestimates costs when several actions are needed to achieve a condition
  - overestimates costs when multiple goals may be achieved with a single action
- **Choosing a good precondition** to refine has also a strong impact
  - select open condition that can be satisfied in the fewest number of ways
    - analogous to most-constrained variable heuristic from CSP
  - Two important special cases:
    - select a condition that cannot be achieved at all (early failure!)
    - select deterministic conditions that can only be achieved in one way

# Planning Graph

- A **planning graph** is a special structure used to
  - achieve better heuristic estimates.
  - directly extract a solution using GRAPHPLAN algorithm
- Consists of a **sequence of levels** (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.
  - Only a restricted subset of possible negative interactions among actions is recorded
- Planning graphs work only for propositional problems
  - STRIPS and ADL can be propositionalized

# Cake Example

- Initial state: `have (cake)`
- Goal state: `have (cake) , eaten (cake)`

```

Action( eat (cake) ,
PRECOND: have (cake)
ADD:     eaten (cake)
DELETE:  have (cake)
)

```

```

Action( bake (cake) ,
PRECOND: not (have (cake) )
ADD:     have (cake)
DELETE:  -
)

```

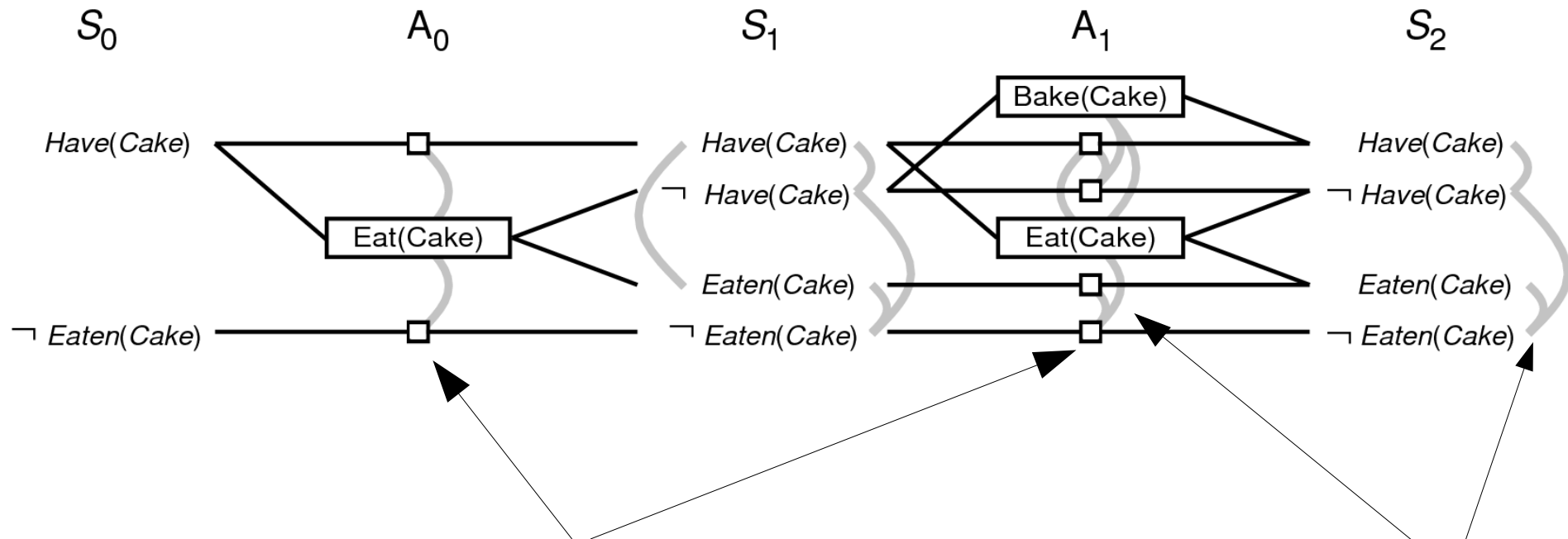
## Persistence Actions

- pseudo-actions for which the effect equals the precondition
- analogous to frame axioms
- are automatically added by the planner

## Mutual exclusions

- link actions or preconditions that are mutually exclusive (*mutex*)

# Cake Example



## Persistence Actions ( $\square$ )

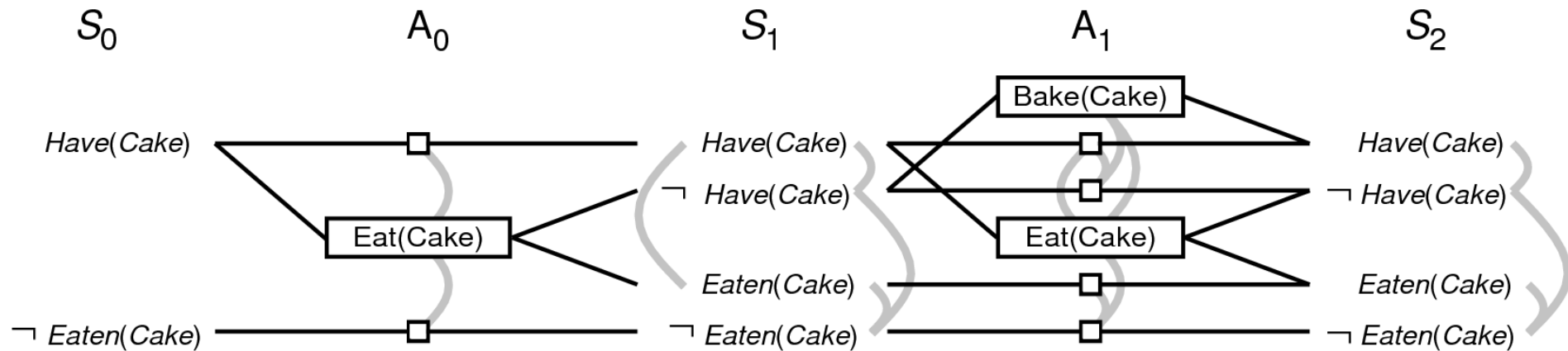
- pseudo-actions for which the effect equals the precondition
- analogous to frame axioms
- are automatically added by the planner

## Mutual exclusions ( — )

- link actions or preconditions that are mutually exclusive (*mutex*)

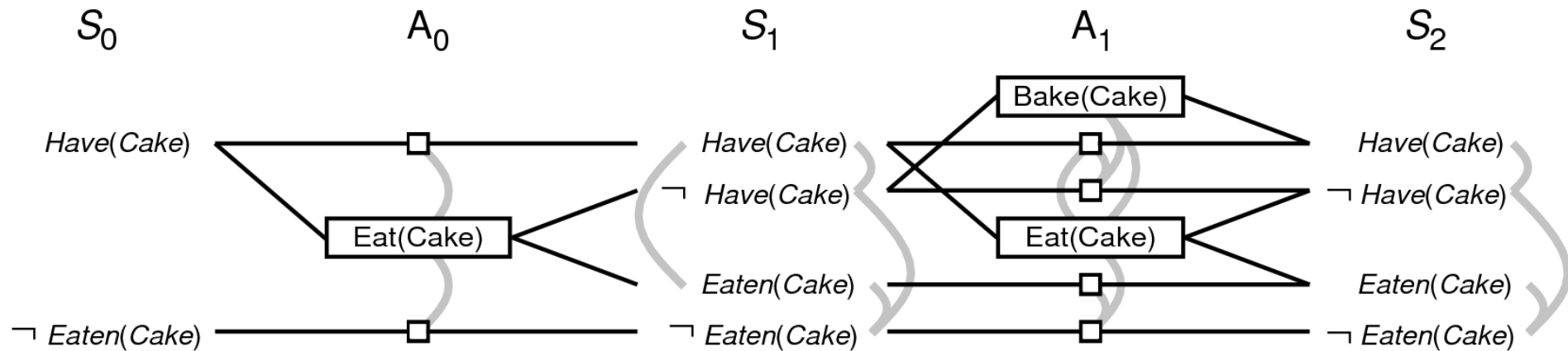


# Cake Example



- Start at level  $S_0$ , determine action level  $A_0$  and next level  $S_1$ 
  - $A_0$  contains all actions whose preconditions are satisfied in the previous level  $S_0$
  - Connect preconditions and effects of these actions
  - Inaction is represented by persistence actions
- Level  $A_0$  contains the actions that could occur
  - Conflicts between actions are represented by mutex links

# Cake Example



- Per construction, 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 are represented by mutex links.
  - $S_1$  defines multiple states and the mutex links are the constraints that define this set of states
- Continue until two consecutive levels are identical
  - Or contain the same amount of literals (explanation later)

# Mutex Relations

- A mutex relation holds between **two actions** when:
  - **Inconsistent effects:**
    - one action negates the effect of another.
  - **Interference:**
    - one of the effects of one action is the negation of a precondition of the other
  - **Competing needs:**
    - one of the preconditions of one action is mutually exclusive with the precondition of the other.
- A mutex relation holds between **two literals** when:
  - **Inconsistent support:**
    - If one is the negation of the other OR
    - if each possible action pair that could achieve the literals is mutex

# Deriving Heuristics from the PG

- Planning Graphs provide information about the problem
  - Example:
    - A literal that does not appear in the final level of the graph cannot be achieved by any plan
- Useful for backward search
  - Any state with an unachievable precondition has cost =  $+\infty$
  - Any plan that contains an unachievable precondition has cost =  $+\infty$
  - In general: **level cost** = level of first appearance of a literal
    - clearly, level cost are an admissible search heuristic
- **Serial Plan Graph**
  - PG allows several actions to occur simultaneously at a level
  - can be serialized by restricting PG to one action per level
    - add mutex links between every pair of actions
  - provides a better heuristic for serial plans
- PG may be viewed a relaxed problem
  - checking only for consistency between pairs of actions/literals

# Costs for Conjunctions of Literals

- **Max-level**: maximum of the goal
  - admissible but not accurate
- **Sum-level**: sum of the level costs
  - makes the subgoal independence assumption
  - inadmissible, but works well in practice
  - Cake Example:
    - estimated costs for **have (cake)  $\wedge$  eaten (cake)** is  $0+1=1$
    - true costs are 2
  - Cake Example without action **bake (cake)**
    - estimated costs are the same
    - true costs are  $+\infty$
- **Set-level**: find the level at which all literals appear and no pair has a mutex link
  - gives the correct estimate in both examples above
  - dominates max-level heuristic, works well with interactions

# The GRAPHPLAN Algorithm

- Algorithm for extracting a solution directly from the PG
  - alternates solution extraction and graph expansion steps

```

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

# GRAPHPLAN Example

- $S_0$  consist of 5 literals (initial state and the CWA literals)

$S_0$

*At(Spare, Trunk)*

*At(Flat, Axle)*

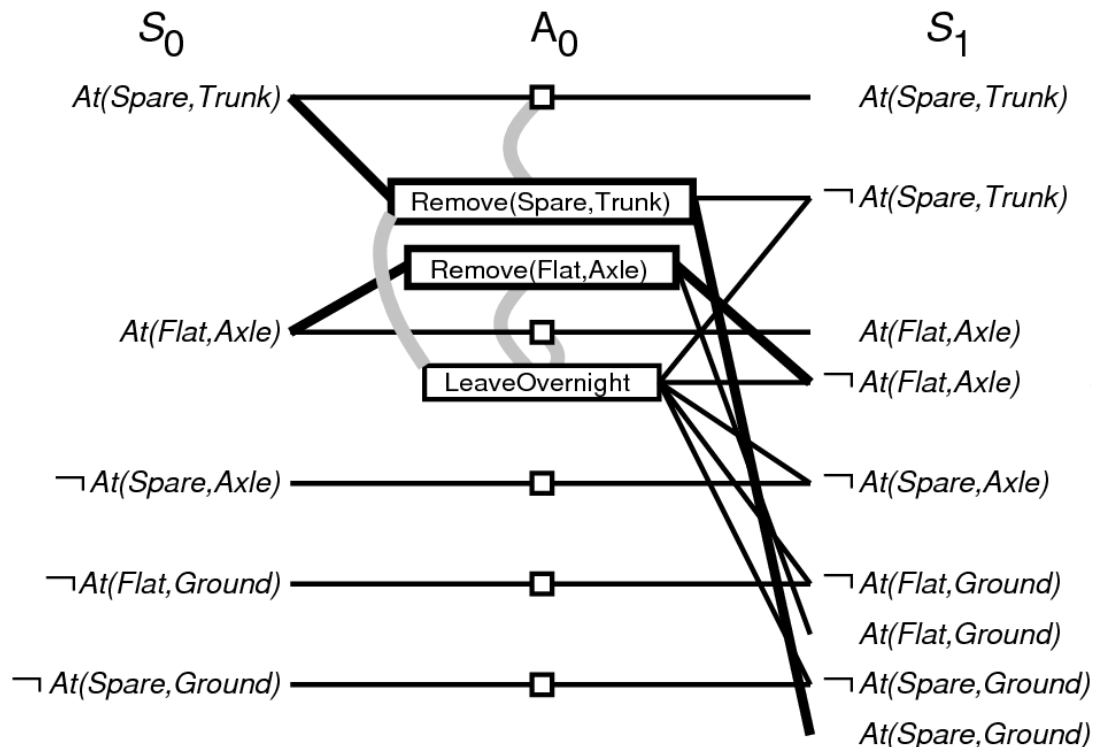
$\neg$ *At(Spare, Axle)*

$\neg$ *At(Flat, Ground)*

$\neg$ *At(Spare, Ground)*

# GRAPHPLAN Example

- $S_0$  consist of 5 literals (initial state and the CWA literals)
- EXPAND-GRAPH adds actions with satisfied preconditions
  - add the effects at level  $S_1$
  - also add persistence actions and mutex relations

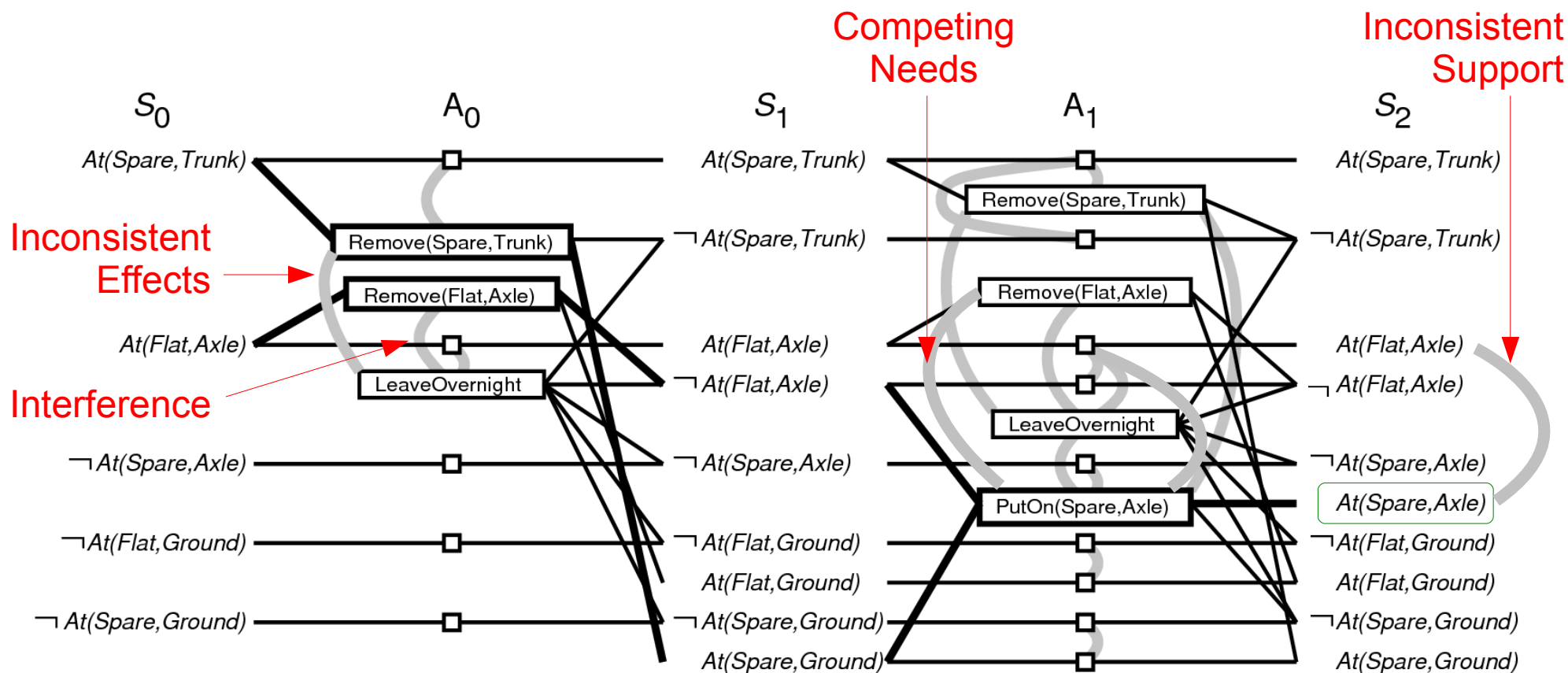




# GRAPHPLAN Example

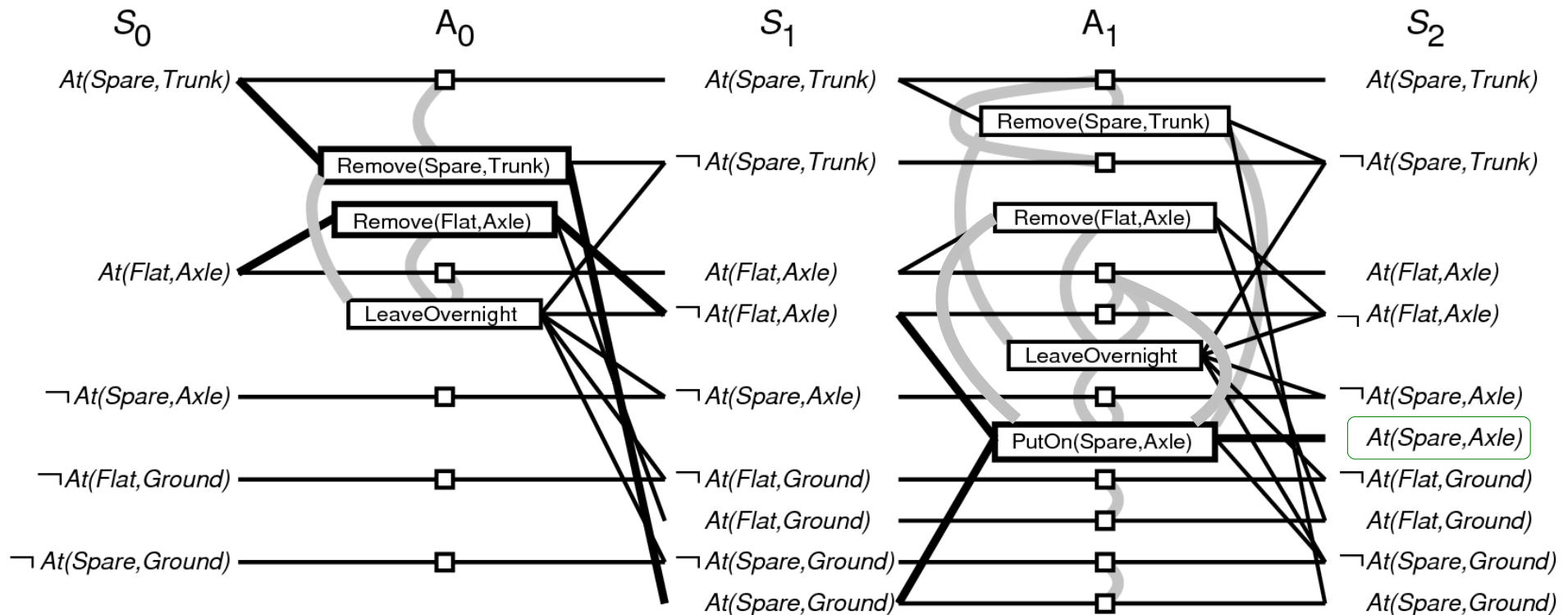
- Repeat

**Note:** Not all mutex links are shown!



# GRAPHPLAN Example

- Repeat until all goal literals are pairwise non-mutex in  $S_i$ 
  - Solution might exist and EXTRACT-SOLUTION will try to find it



# EXTRACT-SOLUTION

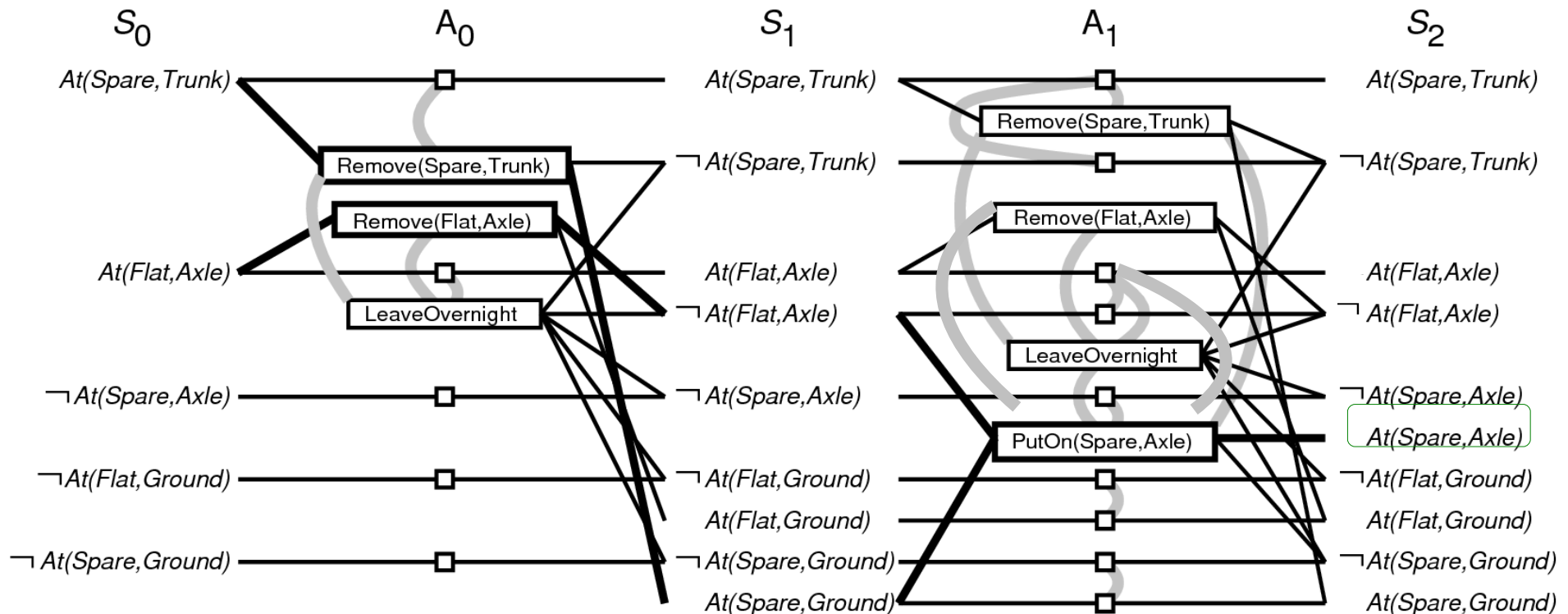
A **state** consists of

- 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
- **Cost**
  - 1 for each action

Could also be formulated as a Boolean CSP

# GRAPHPLAN Example

- Start with goal state  $\text{at}(\text{spare}, \text{axle})$  in  $S_2$
- only action choice is  $\text{puton}(\text{spare}, \text{axle})$  with preconditions  $\text{not}(\text{at}(\text{spare}, \text{axle}))$  and  $\text{at}(\text{spare}, \text{ground})$  in  $S_1$
- $\text{leave-overnight}$  is mutex with  $\text{remove}(\text{spare}, \text{trunk})$   
 $\rightarrow \text{remove}(\text{spare}, \text{trunk})$  and  $\text{remove}(\text{flat}, \text{axle})$
- preconditions are satisfied in  $S_0 \rightarrow$  we're done



# Termination of GRAPHPLAN

1. The planning graph converges because everything is finite
  - number of **literals** is monotonically **increasing**
    - a literal can never disappear because of the persistence actions
  - number of **actions** is monotonically **increasing**
    - once an action is applicable it will always be applicable (because its preconditions will always be there)
  - number of **mutexes** is monotonically **decreasing** for a fixed set of literals
    - more precisely: If two actions are mutex at one level, they are also mutex in all previous levels in which they appear together
    - inconsistent effects and interferences are properties of actions  
→ if they hold once, they will always hold
    - competing needs are properties of mutexes  
→ if the number of actions goes up, chances increase that there is a pair of non-mutex actions that achieve the preconditions
2. It can also be shown that EXTRACT-SOLUTION will find a solution in one of the subsequent expansions of the PG

# SATPLAN

- Key idea:
  - translate the planning problem into **propositional logic**
  - similar to situation calculus, but all facts and rules are ground
    - the same literal in different situations is represented with two different propositions (we call them propositions at a depth  $i$ )
  - actions are also represented as propositions
  - rules are used to derive propositions of depth  $i+1$  from actions and propositions of depth  $i$
- Goal:
  - find a true formula consisting of propositions of the **initial state**, propositions of the **goal state**, and **some action propositions**
- Method:
  - use a satisfiability solver with iterative deepening on the depth
    - first try to prove the goal in depth 0 (initial state)
    - then try to prove the goal in depth 1
    - .... until a solution is found in depth  $n$

**the plan!**

# Key Problem

## ■ Complexity

- In the worst case, a proposition has to be generated
  - for each of  $a$  actions with
  - each of  $o$  possible objects in the  $n$  arguments
  - for a solution depth  $d$
- maximum number of propositions is  $d \cdot a \cdot o^n$
- the number of rules is even larger

## Solution Attempt: Symbol Splitting

- a possible solution is to convert each  $n$ -ary relation into  $n$  binary relation
  - “the  $i$ -th argument of relation  $r$  is  $y$ ”
- this will also reduce the size of the knowledge base because arguments that are not used can be omitted from the rules
- Drawback: multiple instances of the same rule get mixed up
  - no two actions of same type at the same time step
- Nevertheless, SATPLAN is very competitive