

Planning in Artificial Intelligence

The intelligent way to do things

COURSE: CS60045

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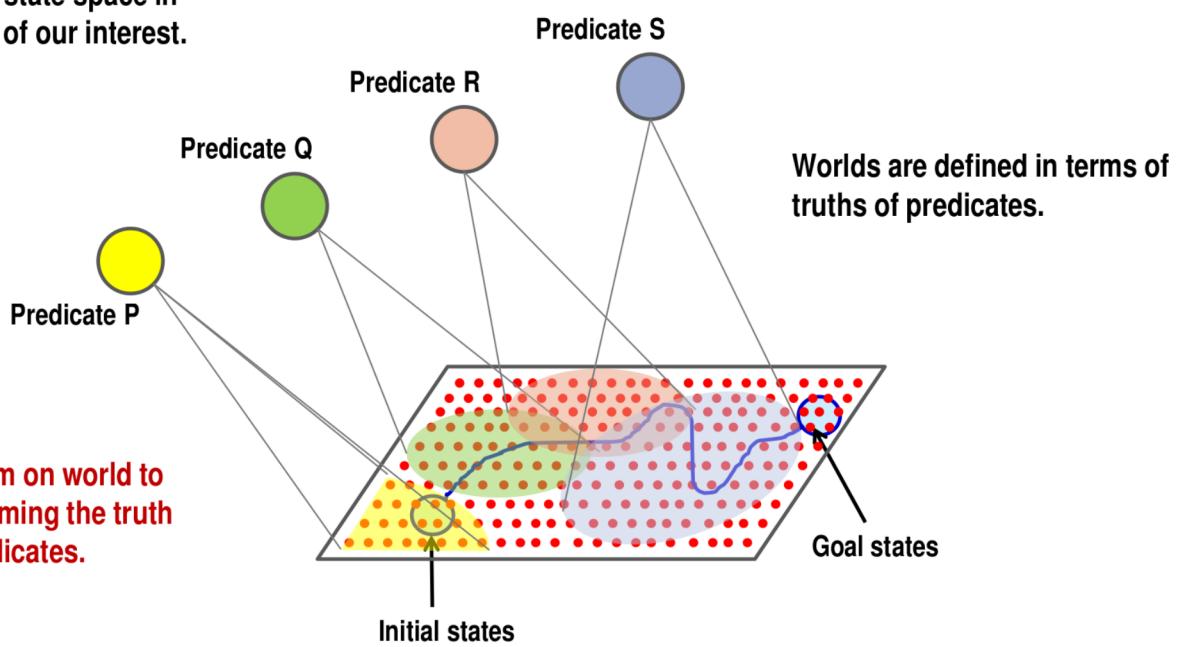


INDIAN INSTITUTE OF TECHNOLOGY KHARAGPUR

From State Spaces to Predicate Worlds

We abstract out the state space in terms of predicates of our interest.

Actions take us from one world to another by transforming the truth of one or more predicates.



Blocks World



Predicates describing the initial state:
On(C, A), On(A, Table),
On(B, Table),
Clear(C), Clear(B)

The planning task is to determine the actions for reaching the target state from the initial state.

Predicates describing the target state:
On(A, B), On(B, C)

ACTIONS:
Move(X, Y)
Precond: Clear(X), Clear(Y)
Effect: On(X, Y)

Move(X, Table)
Precond: Clear(X)
Effect: On(X, Table)

Choosing Actions

ACTIONS:

Move(X, Y)

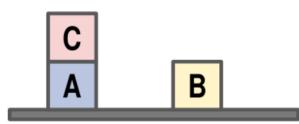
Precond: Clear(X), Clear(Y)

Effect: On(X, Y)

Move(X, Table)

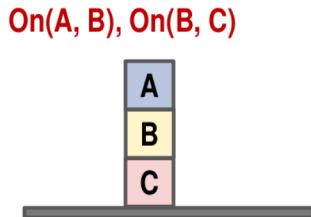
Precond: Clear(X)

Effect: On(X, Table)



On(C, A), On(A, Table), On(B, Table), Clear(C), Clear(B)

- We can move C to the table
 - This achieves none of the goal predicates
- We can move C to top of B
 - This achieves none of the goal predicates
- We can move B to top of C
 - This achieves On(B, C)



Partial Solutions

ACTIONS:

Move(X, Y)

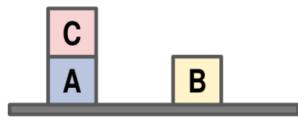
Precond: Clear(X), Clear(Y)

Effect: On(X, Y)

Move(X, Table)

Precond: Clear(X)

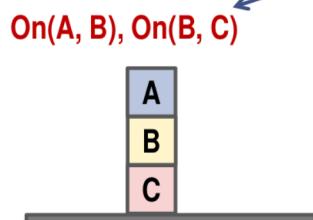
Effect: On(X, Table)



On(C, A), On(A, Table), On(B, Table), Clear(C), Clear(B)

Clear(C), Clear(B)

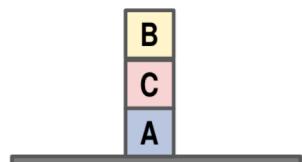
Move(B, C)



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

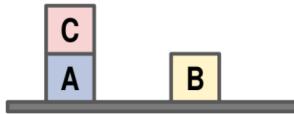
We use Move(B, C) to achieve the sub-goal, On(B, C).

But if we apply this move at the beginning, we get:



Which is not what we want !!

Partial Solutions



On(C, A), On(A, Table), On(B, Table), Clear(C), Clear(B)

Clear(C)

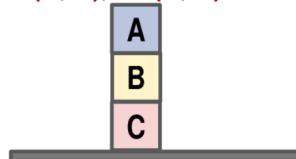
Move(C, Table)

Clear(A), On(C, Table)

Clear(A), Clear(B)

Move(A, B)

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



ACTIONS:

Move(X, Y)

Precond: Clear(X), Clear(Y)

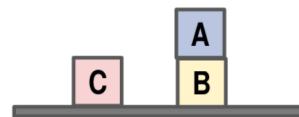
Effect: On(X, Y)

Move(X, Table)

Precond: Clear(X)

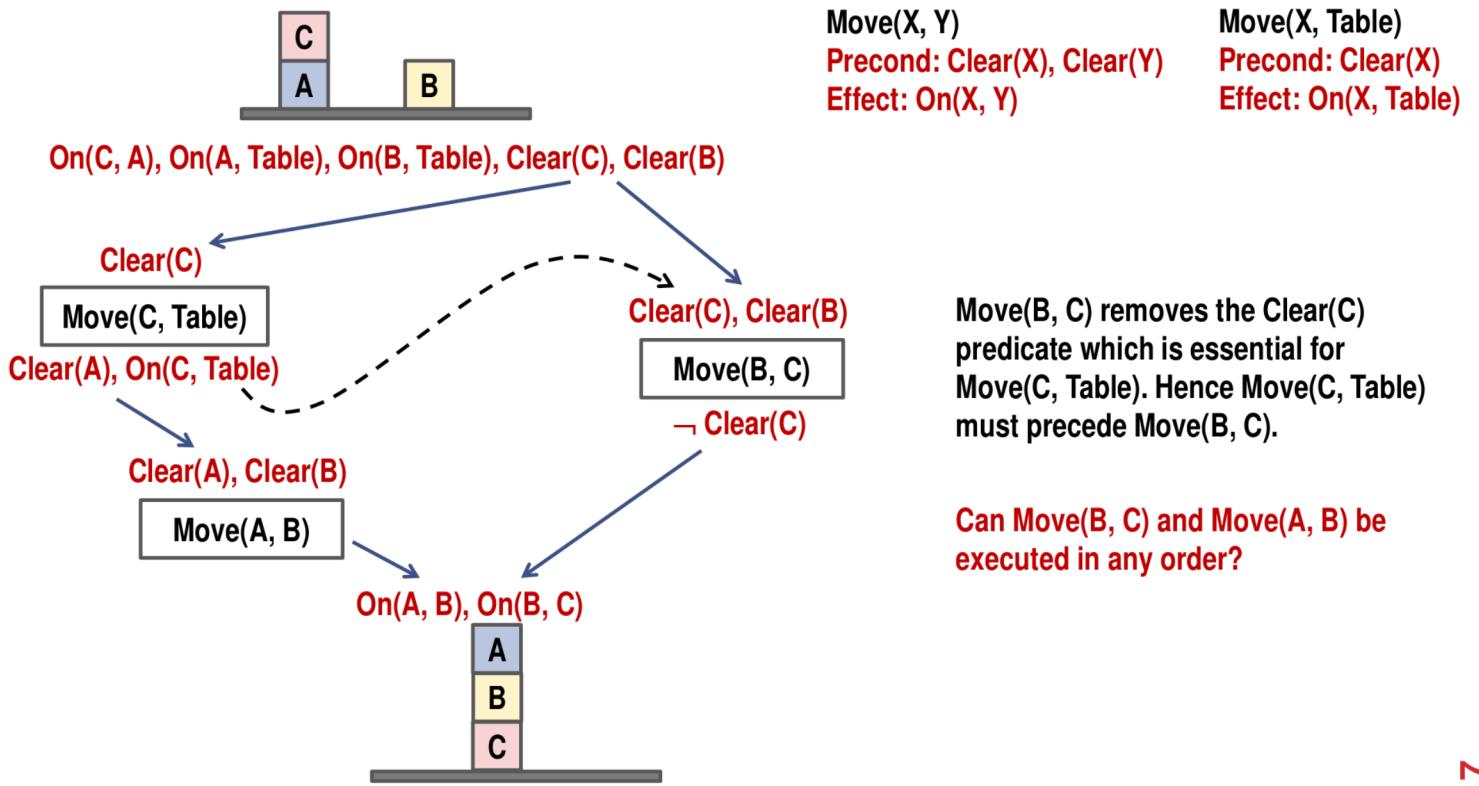
Effect: On(X, Table)

The sub-goal On(A, B) is achieved by moving C to the table and then moving A to top to B. But this gives us:

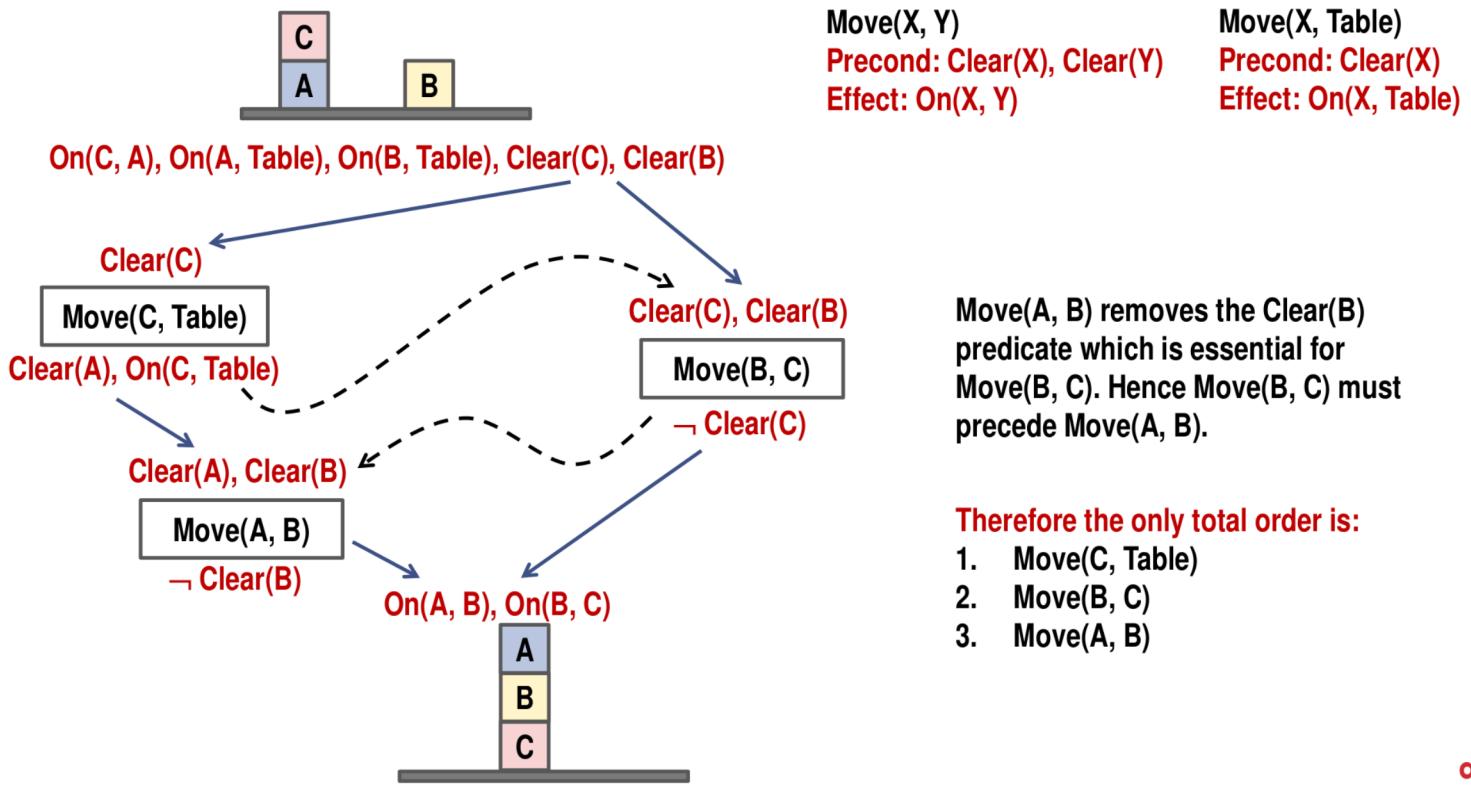


But this too is not what we want !!

Ordering Partial Solutions



Ordering Partial Solutions



Sometimes Partial Order may stay

ACTIONS

Op(**ACTION:** RightShoe,
PRECOND: RightSockOn,
EFFECT: RightShoeOn)

Op(**ACTION:** RightSock,
EFFECT: RightSockOn)

Op(**ACTION:** LeftShoe,
PRECOND: LeftSockOn,
EFFECT: LeftShoeOn)

Op(**ACTION:** LeftSock,
EFFECT: LeftSockOn)

Which of these situations are allowed by these actions?



Sometimes Partial Order may stay

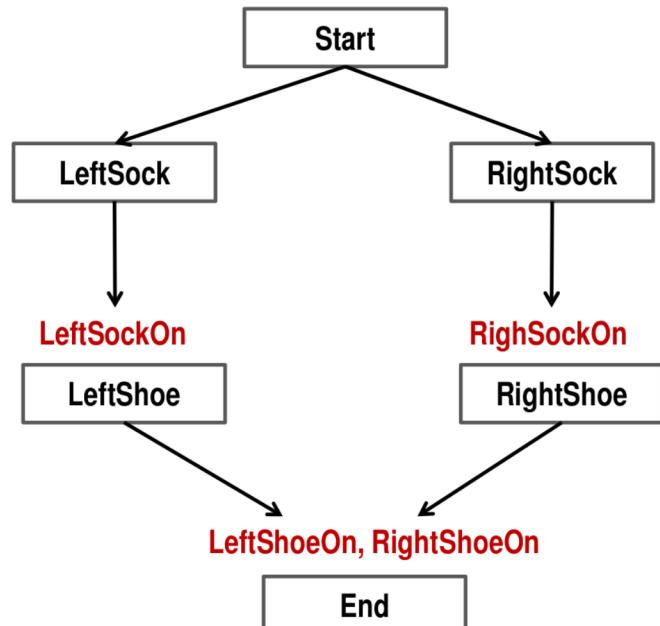
ACTIONS

Op(ACTION: RightShoe,
PRECOND: RightSockOn,
EFFECT: RightShoeOn)

Op(ACTION: RightSock,
EFFECT: RightSockOn)

Op(ACTION: LeftShoe,
PRECOND: LeftSockOn,
EFFECT: LeftShoeOn)

Op(ACTION: LeftSock,
EFFECT: LeftSockOn)



Planning is an integral part of automation

Recommended clip from Charlie Chaplin's Modern Times to see what can go wrong:

https://www.youtube.com/watch?v=n_1apYo6-Ow

What we intend to learn:

1. Partial Order Planning
2. GraphPlan and SATPlan

Partial Order Planning

- **Basic Idea:** Make choices only that are relevant to solving the current part of the problem
- **Least Commitment Choices**
 - **Orderings:** Leave actions unordered, unless they must be sequential
 - **Bindings:** Leave variables unbound, unless needed to unify with conditions being achieved
 - **Actions:** Usually not subject to “least commitment”

Terminology

- **Totally Ordered Plan**
 - There exists sufficient orderings O such that all actions in A are ordered with respect to each other
- **Fully Instantiated Plan**
 - There exists sufficient constraints in B such that all variables are constrained to be equal to some constant
- **Consistent Plan**
 - There are no contradictions in O or B
- **Complete Plan**
 - Every precondition P of every action A_i in A is achieved:
 - There exists an effect of an action A_j that comes before A_i and unifies with P , and no action A_k that deletes P comes between A_j and A_i

Early Days: STRIPS

- STanford Research Institute Problem Solver
- Many planners today use specification languages that are variants of the one used in STRIPS

Our running example:

- Given:
 - **Initial state:** The agent is at *home* without tea, biscuits, book
 - **Goal state:** The agent is at *home* with tea, biscuits, book
 - A set of actions as shown next

Representing States

- States are represented by conjunctions of function-free ground literals

$\text{At(Home)} \wedge \neg \text{Have(Tea)} \wedge$
 $\neg \text{Have(Biscuits)} \wedge \neg \text{Have(Book)}$

- Goals are also described by conjunctions of literals

$\text{At(Home)} \wedge \text{Have(Tea)} \wedge$
 $\text{Have(Biscuits)} \wedge \text{Have(Book)}$

- Goals can also contain variables

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

- The above goal is *being at a shop that sells tea*

Representing Actions

- Action description – serves as a name
- Precondition – a conjunction of positive literals (why positive?)
- Effect – a conjunction of literals (+ve or -ve)
 - The original version had an *add list* and a *delete list*.

Op(ACTION: Go(there),
PRECOND: At(here) \wedge Path(here, there),
EFFECT: At(there) \wedge \neg At(here))

Representing Plans

- A set of plan steps. Each step is one of the operators for the problem.
- A set of step ordering constraints. Each ordering constraint is of the form $S_i \prec S_j$, indicating S_i must occur sometime before S_j .
- A set of variable binding constraints of the form $v = x$, where v is a variable in some step, and x is either a constant or another variable.
- A set of causal links written as $S \rightarrow c: S'$ indicating S satisfies the precondition c for S' .

Example

- Initial plan

Plan(

STEPS: {

S1: Op(ACTION: start),

S2: Op(ACTION: finish,

PRECOND: RightShoeOn \wedge LeftShoeOn)

},

ORDERINGS: { $S_1 \prec S_2$ },

BINDINGS: { },

LINKS: { })

POP Example: Get Tea, Biscuits, Book

Initial state:

Op(**ACTION:** Start,
EFFECT: At(Home) \wedge Sells(BS, Book)
 \wedge Sells(TS, Tea)
 \wedge Sells(TS, Biscuits))

Goal state:

Op(**ACTION:** Finish,
PRECOND: At(Home) \wedge Have(Tea)
 \wedge Have(Biscuits)
 \wedge Have(Book))

Actions:

Op(**ACTION:** Go(y),
PRECOND: At(x),
EFFECT: At(y) \wedge \neg At(x))

Op(**ACTION:** Buy(x),
PRECOND: At(y) \wedge Sells(y, x),
EFFECT: Have(x))

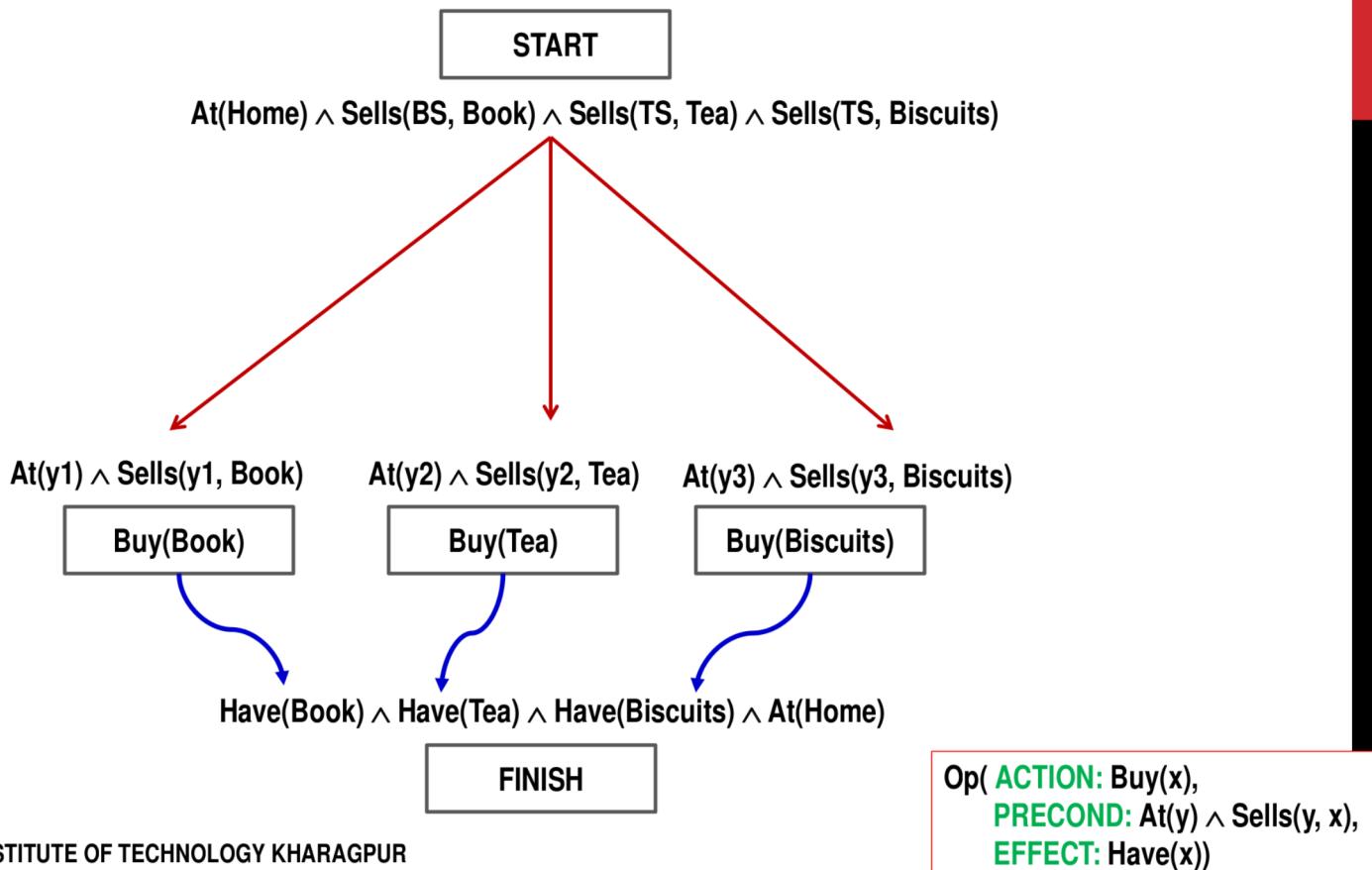
START

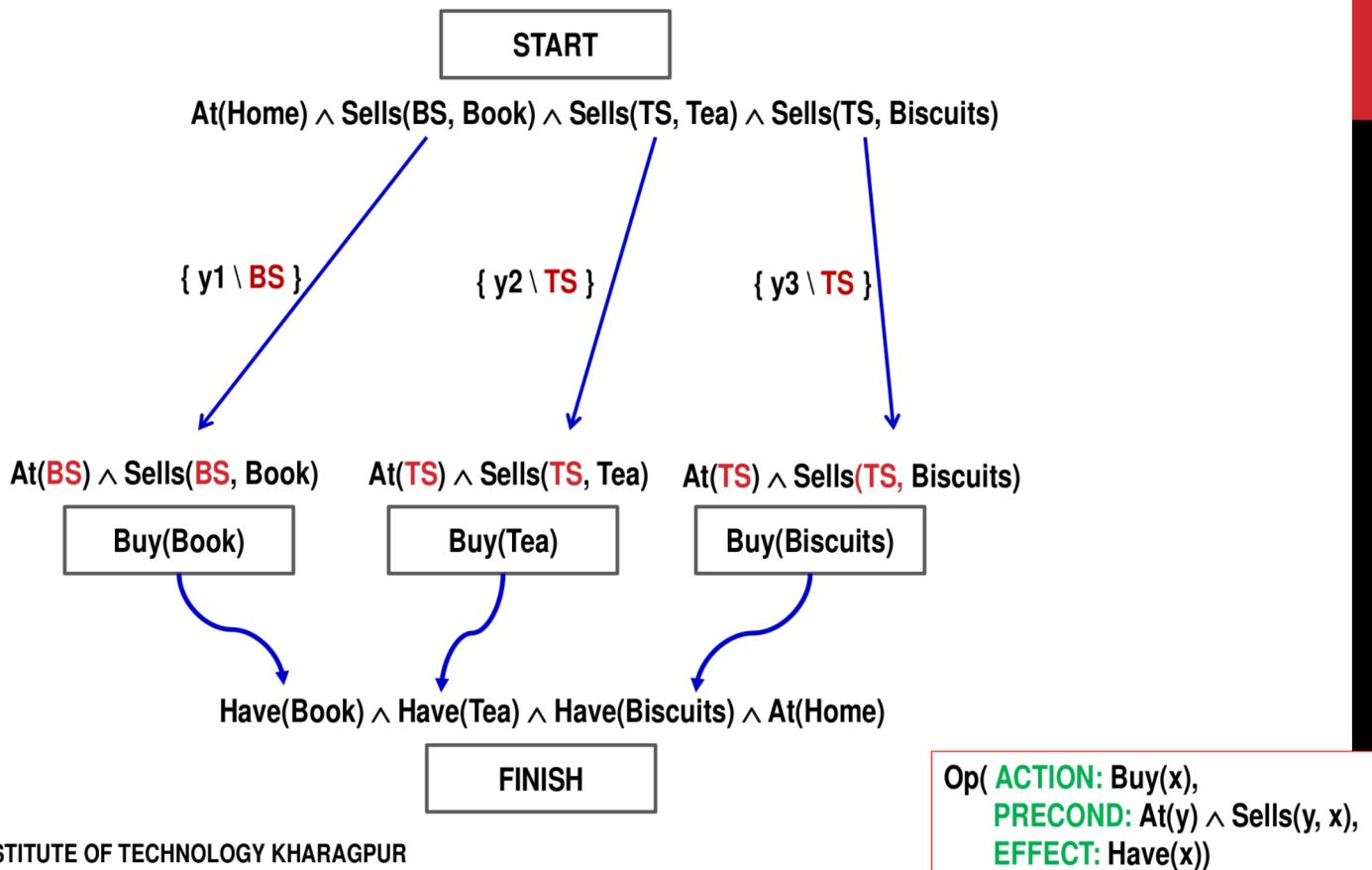
At(Home) \wedge Sells(BS, Book) \wedge Sells(TS, Tea) \wedge Sells(TS, Biscuits)

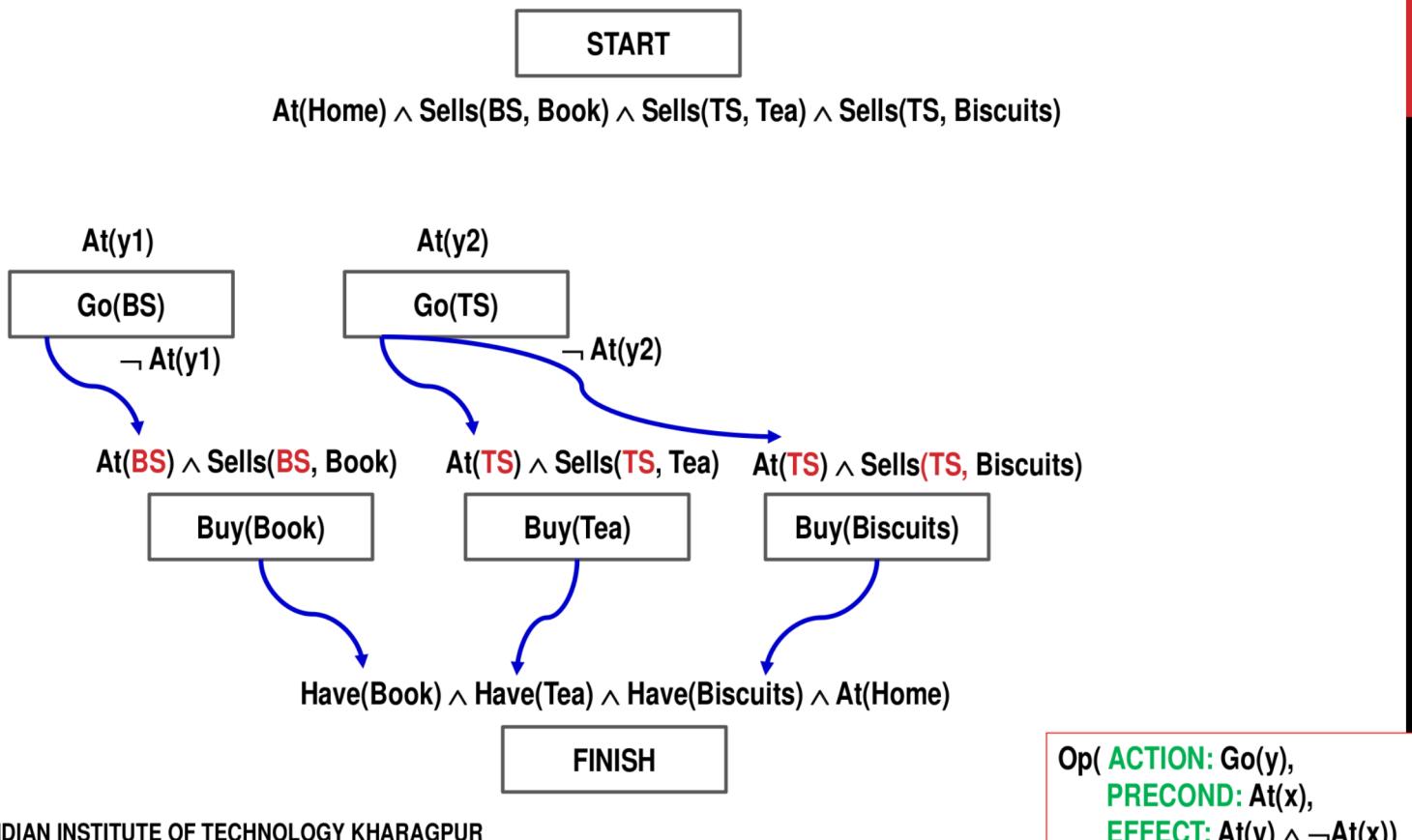


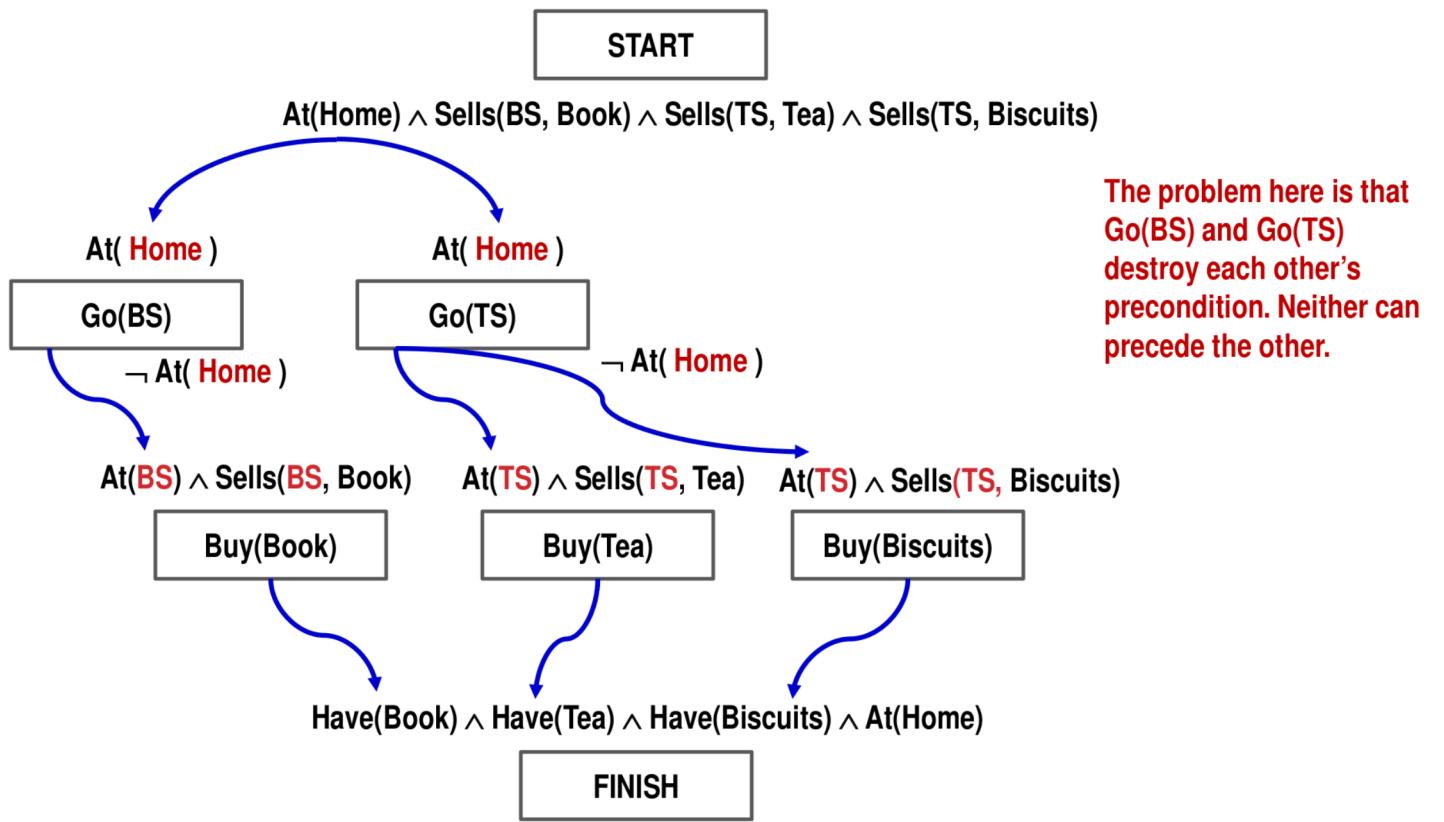
Have(Book) \wedge Have(Tea) \wedge Have(Biscuits) \wedge At(Home)

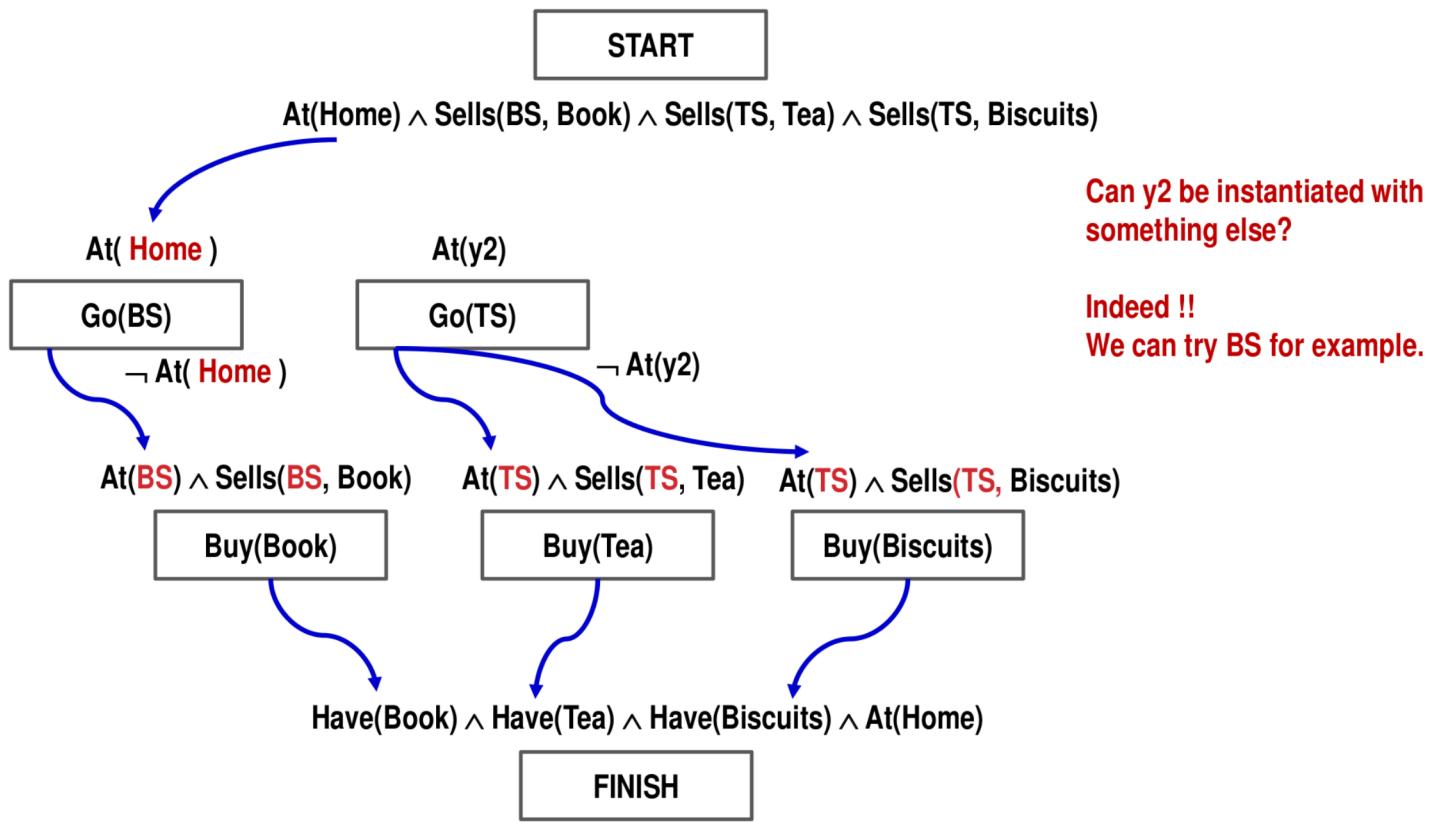
FINISH

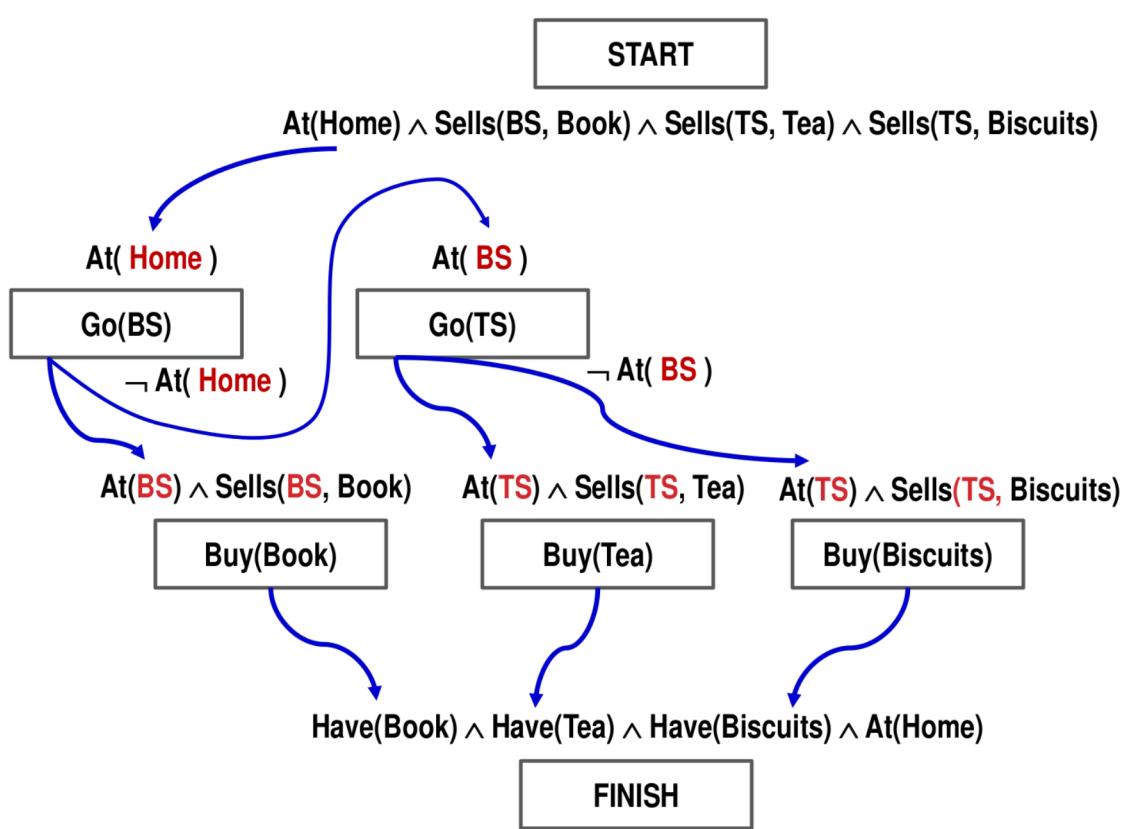


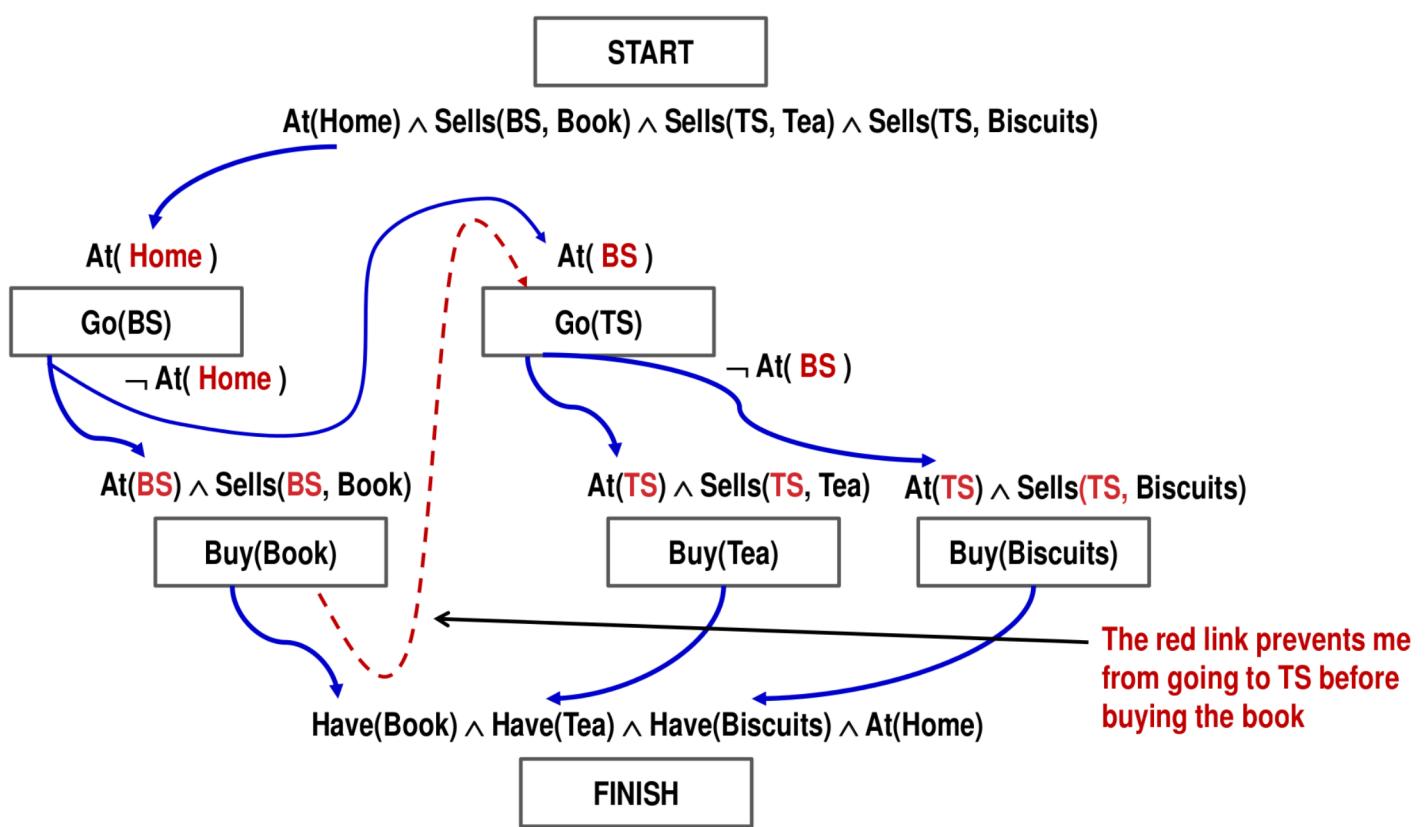


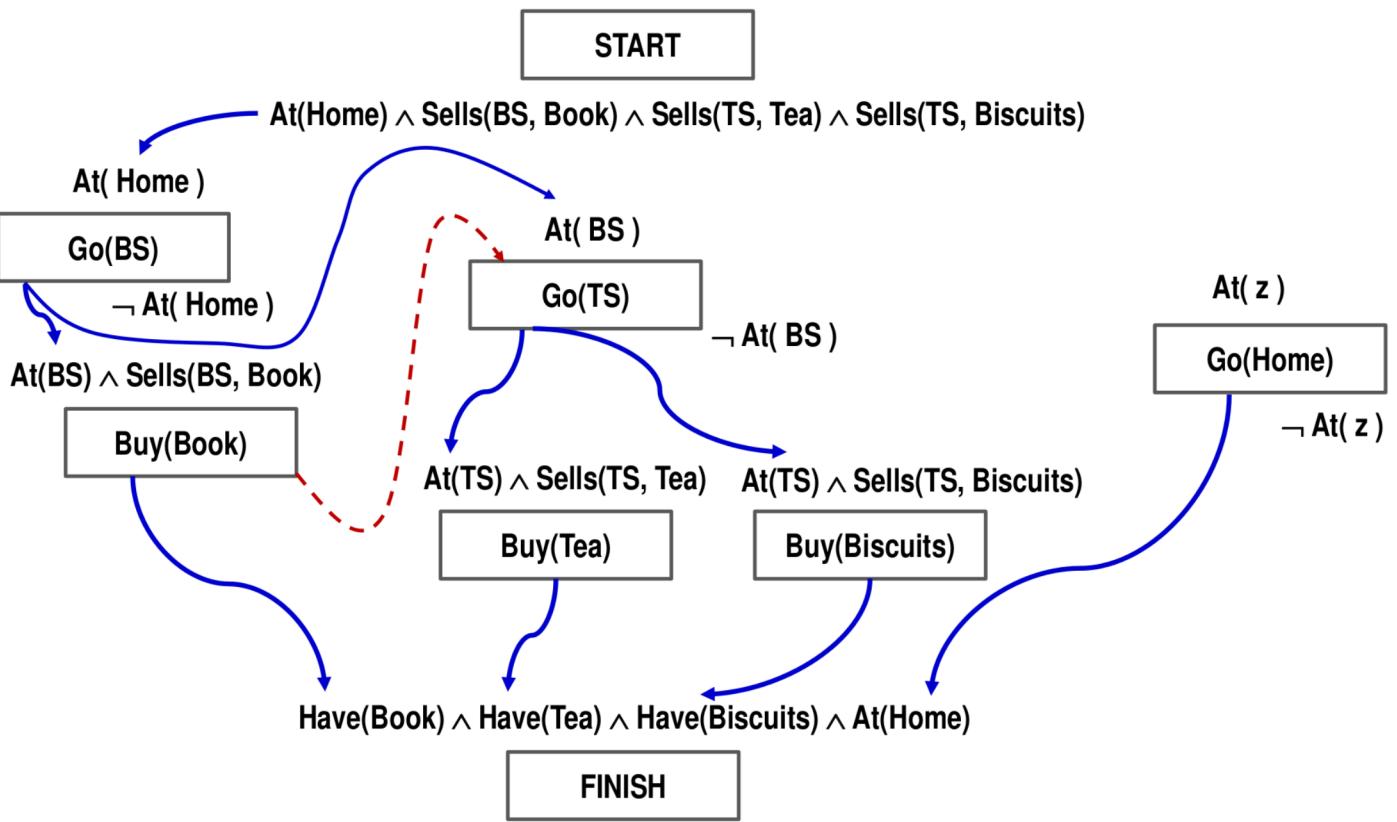


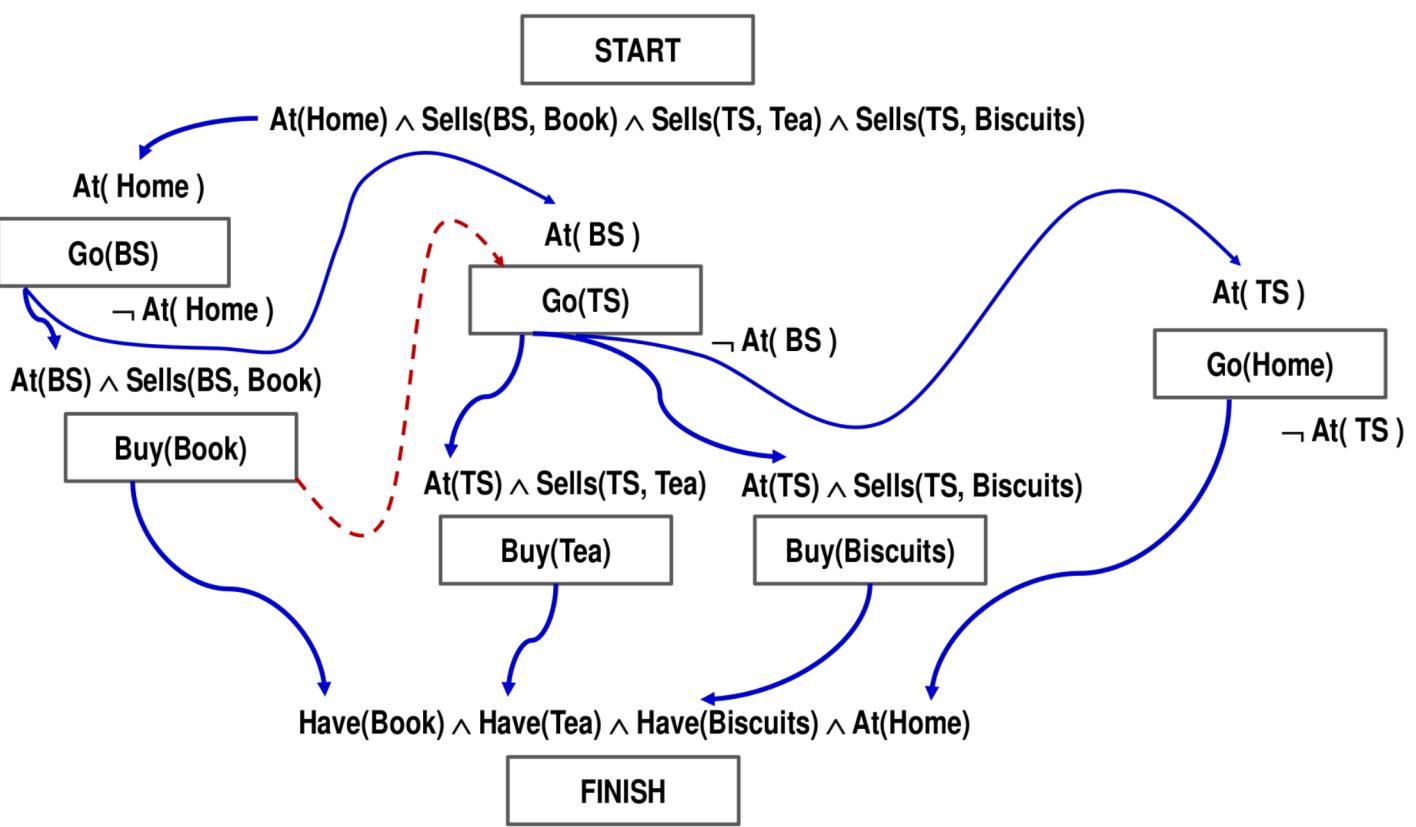


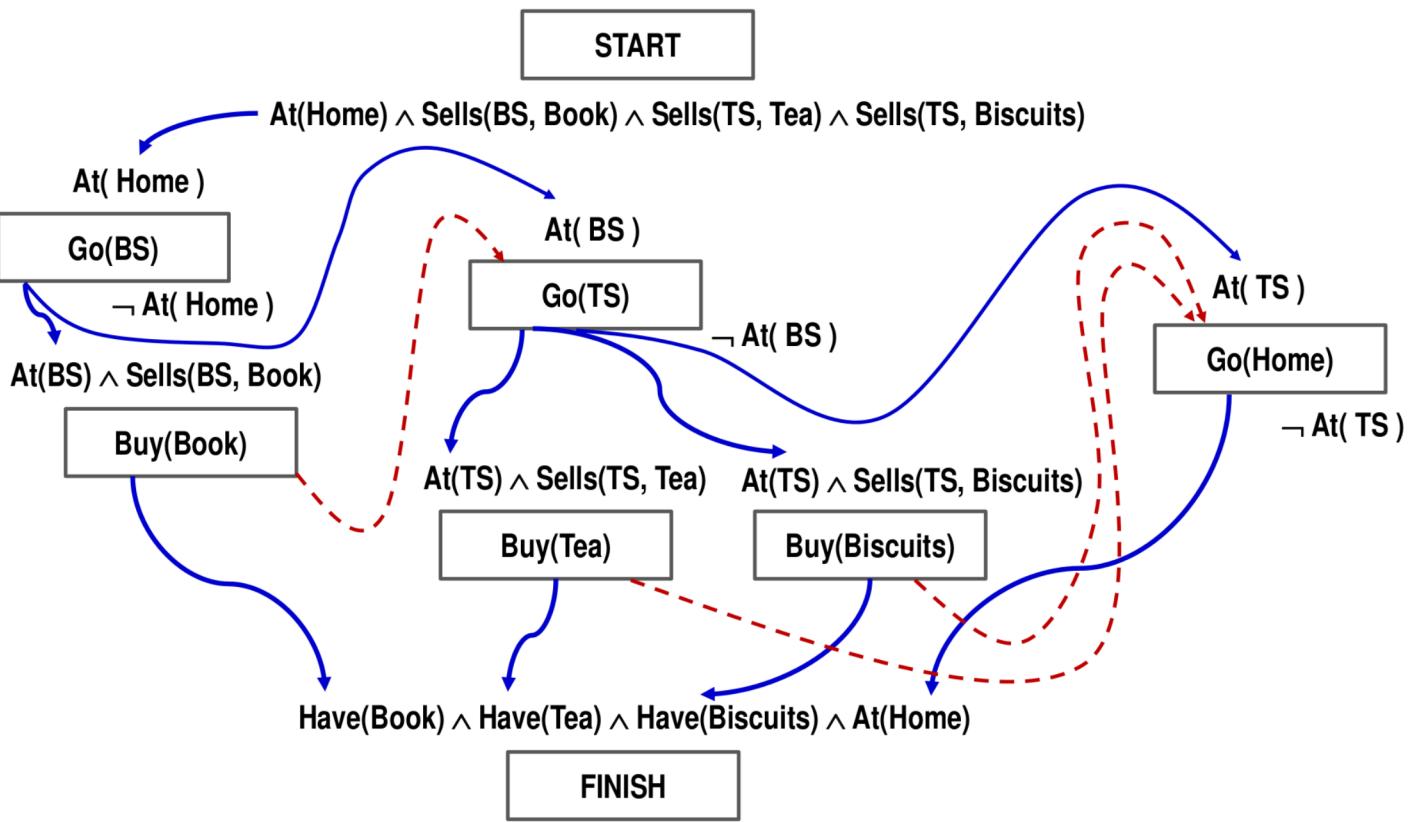


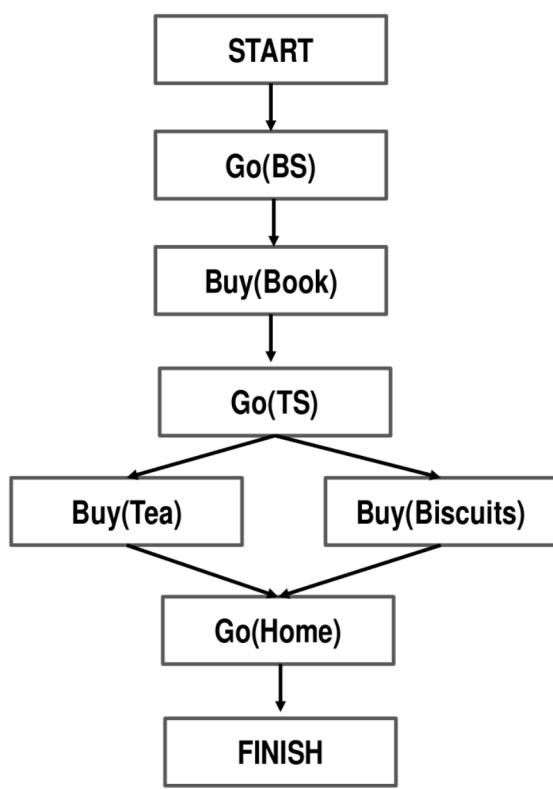












The Partial Order Planning Algorithm

```
Function POP( initial, goal, operators )
// Returns plan
    plan ← Make-Minimal-Plan( initial, goal )
    Loop do
        If Solution( plan ) then return plan
        S, c ← Select-Subgoal( plan )
        Choose-Operator( plan, operators, S, c )
        Resolve-Threats( plan )
    end
```

POP: Selecting Sub-Goals

Function **Select-Subgoal(** *plan* **)**

// Returns S, c

pick a plan step S from STEPS(*plan* **)**

with a precondition C that has not been achieved

Return S, c

POP: Choosing operators

Procedure Choose-Operator(*plan, operators, S, c*)

Choose a step S' from *operators* or $\text{STEPS}(\textit{plan})$ that has c as an effect

If there is no such step then **fail**

Add the causal link $S' \rightarrow c: S$ to $\text{LINKS}(\textit{plan})$

Add the ordering constraint $S' \prec S$ to $\text{ORDERINGS}(\textit{plan})$

If S' is a newly added step from *operators* then add S' to $\text{STEPS}(\textit{plan})$ and add Start $\prec S' \prec$ Finish to $\text{ORDERINGS}(\textit{plan})$

POP: Resolving Threats

Procedure **Resolve-Threats(*plan*)**

for each **S'** that threatens a link $S_i \rightarrow c: S_j$ in **LINKS(*plan*)** do

choose either

Promotion: Add $S'' \prec S_i$ to **ORDERINGS(*plan*)**

Demotion: Add $S_j \prec S''$ to **ORDERINGS(*plan*)**

if not **Consistent(*plan*)** then **fail**

Partially instantiated operators

- So far we have not mentioned anything about binding constraints
- Should an operator that has the effect, say, $\neg At(x)$, be considered a threat to the condition, $At(Home)$?
 - Indeed it is a *possible threat* because x may be bound to $Home$

Dealing with potential threats

Resolve now with an equality constraint

- Bind x to something that resolves the threat (say $x = TS$)

Resolve now with an inequality constraint

- Extend the language of variable binding to allow $x \neq Home$

Resolve later

- Ignore possible threats. If $x = Home$ is added later into the plan, then we will attempt to resolve the threat (by promotion or demotion)

Proc Choose-Operator(*plan*, *operators*, *S*, *c*)

choose a step *S'* from *operators* or STEPS(*plan*) that has *c'* as an effect
such that *u* = UNIFY(*c*, *c'*, BINDINGS(*plan*))

if there is no such step then fail

add *u* to BINDINGS(*plan*)

add the causal link *S' → c*: *S* to LINKS(*plan*)

add the ordering constraint *S' < S* to ORDERINGS(*plan*)

if *S'* is a newly added step from *operators* then

 add *S'* to STEPS(*plan*) and add Start < *S'* < Finish to ORDERINGS(*plan*)

Procedure Resolve-Threats(*plan*)

```
for each  $S_i \rightarrow c: S_j$  in LINKS( plan ) do
    for each  $S''$  in STEPS( plan ) do
        for each  $c'$  in EFFECTS(  $S''$  ) do
            if SUBST( BINDINGS(plan),  $c$  ) = SUBST( BINDINGS(plan),  $\neg c'$  )
                then choose either
                    Promotion: Add  $S'' \prec S_i$  to ORDERINGS( plan )
                    Demotion: Add  $S_i \prec S''$  to ORDERINGS( plan )
            if not Consistent( plan ) then fail
```

USING PLANNING GRAPHS

GraphPlan and SATPlan

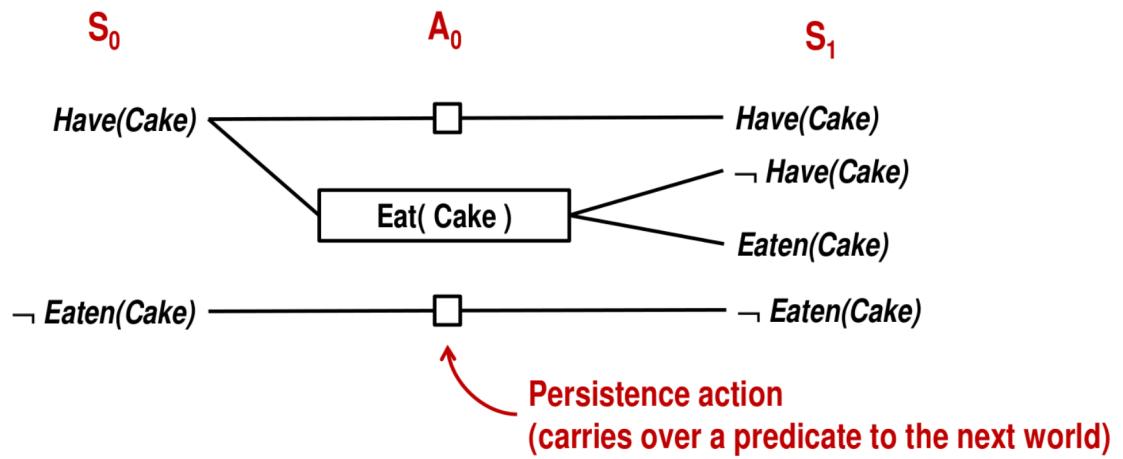
Planning Graph

Start: $\text{Have}(\text{Cake})$

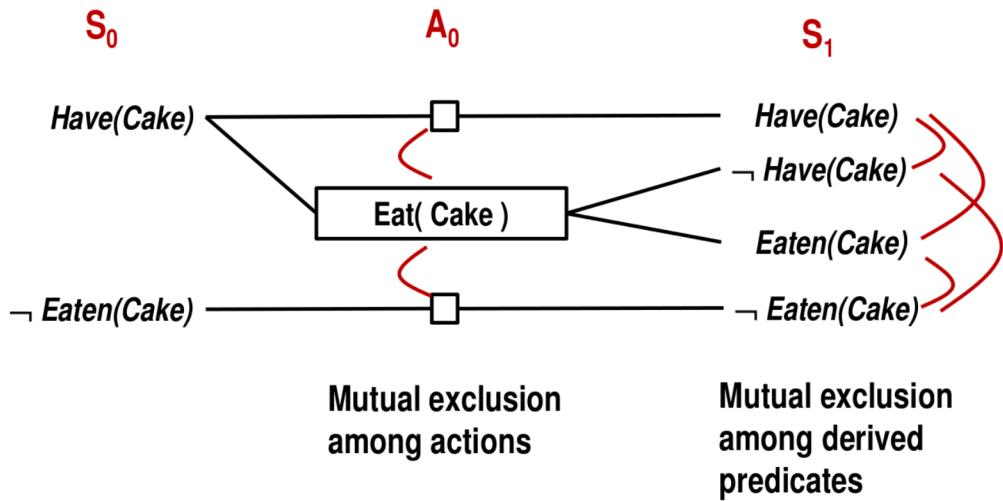
Finish: $\text{Have}(\text{Cake}) \wedge \text{Eaten}(\text{Cake})$

$\text{Op}(\text{ACTION: Eat}(\text{Cake}),$
 $\text{PRECOND: Have}(\text{Cake}),$
 $\text{EFFECT: Eaten}(\text{Cake}) \wedge \neg \text{Have}(\text{Cake}))$

$\text{Op}(\text{ACTION: Bake}(\text{Cake}),$
 $\text{PRECOND: } \neg \text{Have}(\text{Cake}),$
 $\text{EFFECT: Have}(\text{Cake}))$



Mutex Links in a Planning Graph



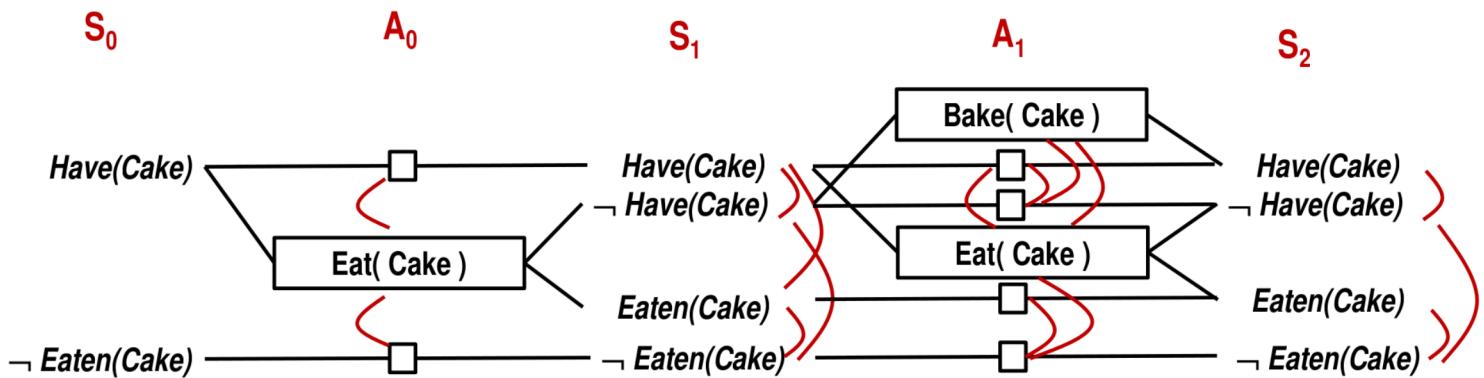
Planning Graphs

- Consists of a sequence of levels that correspond to time steps in the plan
- Each level contains a set of actions and a set of literals that *could* be true at that time step depending on the actions taken in previous time steps
- For every +ve and -ve literal C, we add a *persistence action* with precondition C and effect C

Planning Graph

Op(ACTION: Eat(Cake),
PRECOND: Have(Cake),
EFFECT: Eaten(Cake) $\wedge \neg$ Have(Cake))

Op(ACTION: Bake(Cake),
PRECOND: \neg Have(Cake),
EFFECT: Have(Cake))

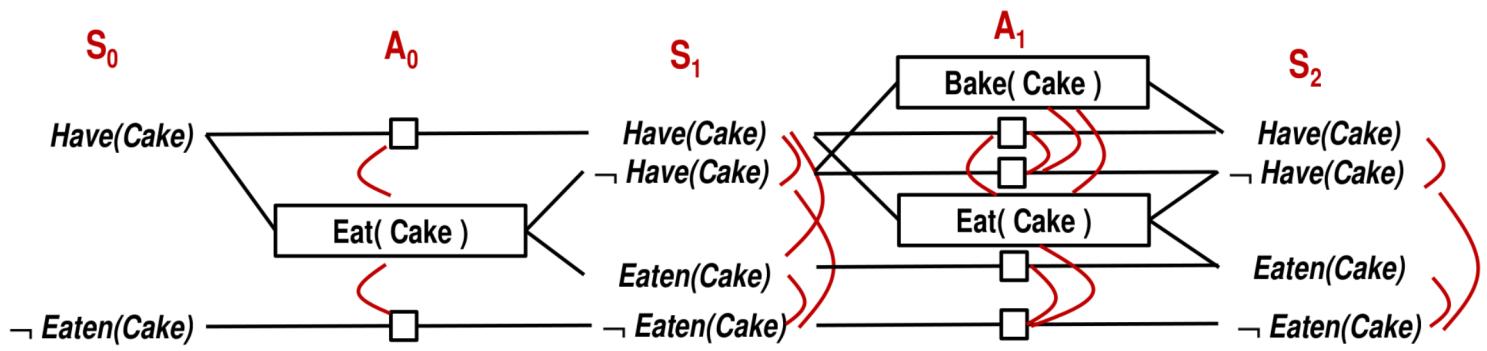


Start: $\text{Have}(\text{Cake})$
Finish: $\text{Have}(\text{Cake}) \wedge \text{Eaten}(\text{Cake})$

In the world S_2 the goal predicates exist without mutexes, hence we need not expand the graph any further

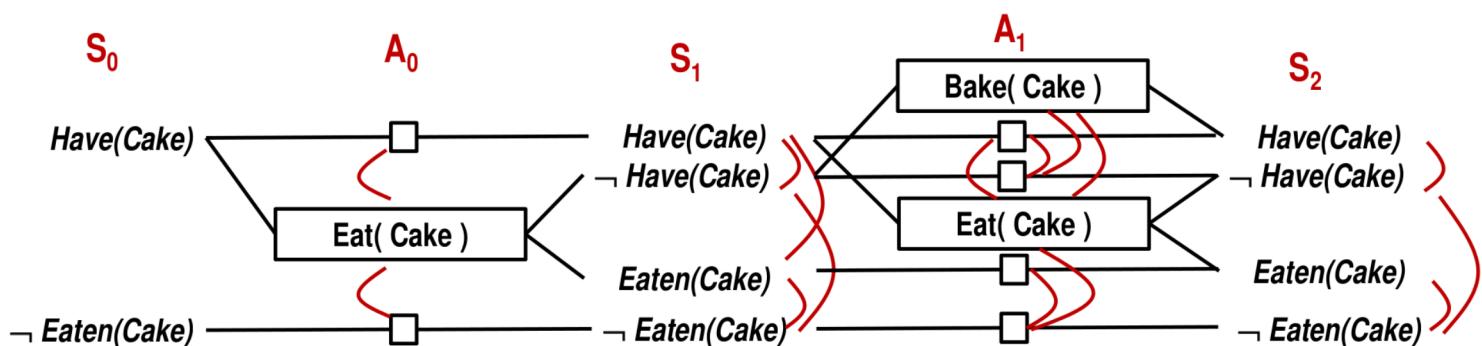
Mutex Actions

- Mutex relation exists between two actions if:
 - **Inconsistent effects** – one action negates an effect of the other
 Eat(Cake) causes \neg Have(Cake) and Bake(Cake) causes Have(Cake)
 - **Interference** – one of the effects of one action is the negation of a precondition of the other
 Eat(Cake) causes \neg Have(Cake) and the persistence of Have(Cake) needs Have(Cake)
 - **Competing needs** – one of the preconditions of one action is mutually exclusive with a precondition of the other
 Bake(Cake) needs \neg Have(Cake) and Eat(Cake) needs Have(Cake)



Mutex Literals

- Mutex relation exists between two literals if:
 - One is the negation of the other, or
 - Each possible pair of actions that could achieve the two literals is mutually exclusive (inconsistent support)

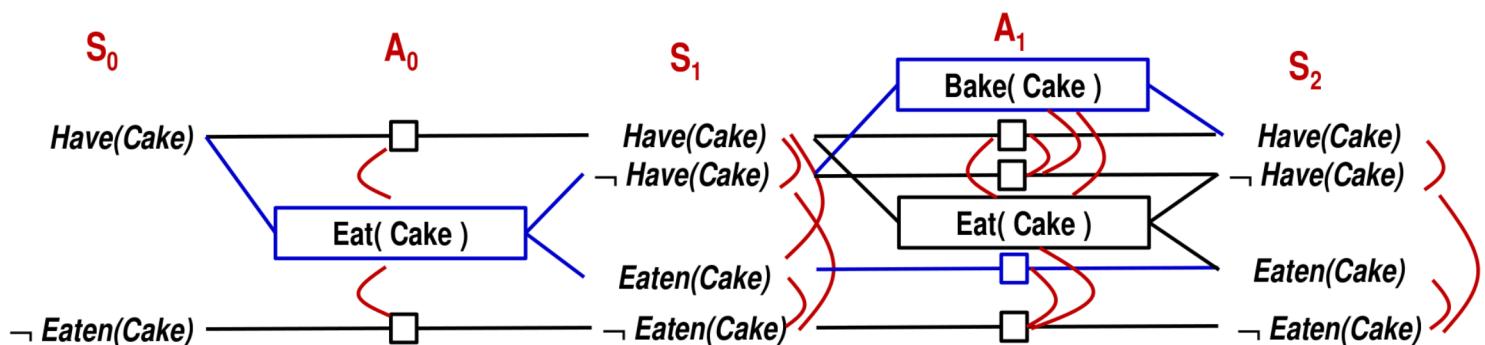


Function GraphPLAN(problem)

```
// returns solution or failure
graph ← Initial-Planning-Graph( problem )
goals ← Goals[ problem ]
do
    if goals are all non-mutex in last level of graph then do
        solution ← Extract-Solution( graph )
        if solution ≠ failure then return solution
        else if No-Solution-Possible (graph )
            then return failure
    graph ← Expand-Graph( graph, problem )
```

Finding the plan

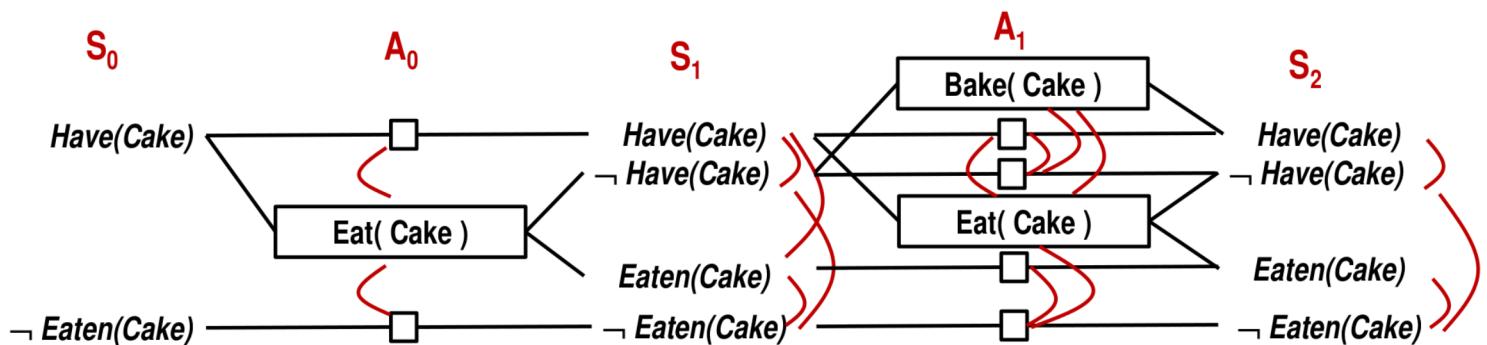
- Once a world is found having all goal predicates without mutexes, the plan can be extracted by solving a constraint satisfaction problem (CSP) for resolving the mutexes
 - Creating the planning graph can be done in polynomial time, but planning is known to be a PSPACE-complete problem. The hardness is in the CSP.
 - The plan is shown in blue below



Termination of GraphPLAN when no plan exists

- Literals increase monotonically
- Actions increase monotonically
- Mutexes decrease monotonically

This guarantees the existence of a fixpoint



Exercise

Start: At(Flat, Axle) \wedge At(Spare, Trunk)

Goal: At(Spare, Axle)

Op(**ACTION:** Remove(Spare, Trunk),
PRECOND: At(Spare, Trunk),
EFFECT: At(Spare, Ground)
 $\wedge \neg$ At(Spare, Trunk))

Op(**ACTION:** Remove(Flat, Axle),
PRECOND: At(Flat, Axle),
EFFECT: At(Flat, Ground)
 $\wedge \neg$ At(Flat, Axle))

Op(**ACTION:** PutOn(Spare, Axle),
PRECOND: At(Spare, Ground)
 $\wedge \neg$ At(Flat, Axle),
EFFECT: At(Spare, Axle)
 $\wedge \neg$ At(Spare, Ground))

Op(**ACTION:** LeaveOvernight,
PRECOND:
EFFECT: \neg At(Spare, Ground)
 $\wedge \neg$ At(Spare, Axle)
 $\wedge \neg$ At(Spare, Trunk)
 $\wedge \neg$ At(Flat, Ground)
 $\wedge \neg$ At(Flat, Axle))

Planning with Propositional Logic

- The planning problem is translated into a CNF satisfiability problem
- The goal is asserted to hold at a time step T, and clauses are included for each time step up to T.
- If the clauses are satisfiable, then a plan is extracted by examining the actions that are true.
- Otherwise, we increment T and repeat

Example

Aeroplanes P_1 and P_2 are at SFO and JFK respectively. We want P_1 at JFK and P_2 at SFO

Initial: $\text{At}(P_1, \text{SFO})^0 \wedge \text{At}(P_2, \text{JFK})^0$

Goal: $\text{At}(P_1, \text{JFK}) \wedge \text{At}(P_2, \text{SFO})^0$

Action: $\text{At}(P_1, \text{JFK})^1 \Leftrightarrow [\text{At}(P_1, \text{JFK})^0 \wedge \neg (\text{Fly}(P_1, \text{JFK}, \text{SFO})^0 \wedge \text{At}(P_1, \text{JFK})^0)]$
 $\quad \vee [\text{At}(P_1, \text{SFO})^0 \wedge \text{Fly}(P_1, \text{SFO}, \text{JFK})^0]$

Check the satisfiability of:

initial state \wedge successor state axioms \wedge goal

Additional Axioms

Precondition Axioms:

$$\text{Fly}(P_1, \text{JFK}, \text{SFO})^0 \Rightarrow \text{At}(P_1, \text{JFK})^0$$

Action Exclusion Axioms:

$$\neg (\text{Fly}(P_2, \text{JFK}, \text{SFO})^0 \wedge \text{Fly}(P_2, \text{JFK}, \text{LAX})^0)$$

State Constraints:

$$\forall p, x, y, t \ (x \neq y) \Rightarrow \neg (\text{At}(p, x)^t \wedge \text{At}(p, y)^t)$$

SATPlan

```
Function SATPlan( problem, Tmax )
    // returns solution or failure

    for T = 0 to Tmax do
        cnf, mapping ← Trans-to-SAT(problem, T)
        assignment ← SAT-Solver( cnf )
        if assignment is not NULL then
            return Extract-Solution(assignment, mapping)
    return failure
```

Further Readings

- Heuristic Search Planning
- Planning with Temporal Goals
- Planning under Adversaries
- Multi-agent Planning
- Planning in Continuous State Spaces
- Planning with Reinforcement Learning

Explainable AI Planning (XAIP)

Enables you to seek explanations from the planner.

- Why did you do that?
- And why didn't you do something else (which I would have chosen)?
- Why is what you propose better / cheaper / safer than what I would have done?
- Why can't you do that?
- Why do I need to backtrack (and replan) at this point?
- Why do I not need to replan at this point?