

Lecture 6b: Search Based Planning

CSCI 360

Introduction to Artificial Intelligence

USC

Here is where we are...

Week	30000D	30282R	Topics	Chapters
1	1/7 1/9	1/8 1/10	Intelligent Agents Problem Solving and Search	[Ch 1.1-1.4 and 2.1-2.4] [Ch 3.1-3.3]
2	1/14 1/16	1/15 1/17	Uninformed Search Heuristic Search (A*)	[Ch 3.3-3.4] [Ch 3.5]
3	1/21 1/23	1/22 1/24	Heuristic Functions Local Search	[Ch 3.6] [Ch 4.1-4.2]
	1/25		Project 1 Out	
4	1/28 1/30	1/29 1/31	Adversarial Search Knowledge Based Agents	[Ch 5.1-5.3] [Ch 7.1-7.3]
5	2/4 2/6	2/5 2/7	Propositional Logic Inference First-Order Logic	[Ch 7.4-7.5] [Ch 8.1-8.4]
	2/8 2/8		Project 1 Due Homework 1 Out	
6	2/11 2/13	2/12 2/14	Rule-Based Systems Search-Based Planning	[Ch 9.3-9.4] [Ch 10.1-10.3]
	2/15		Homework 1 Due	
7	2/18 2/20	2/19 2/21	SAT-Based Planning Knowledge Representation	[Ch 10.4] [Ch 12.1-12.5]
8	2/25 2/27	2/26 2/28	Midterm Review Midterm Exam	



Outline

- What is AI?
- Problem-solving agent
 - Uninformed (DFS), informed (A^*), and local search
 - Adversarial search (minimax, alpha-beta pruning)
- **Knowledge-based agent**
 - Propositional Logic
 - First Order Logic (FOL)
 - **Automated Reasoning in FOL**
 - Substitution
 - Unification (GMP)
 - Chaining (forward and backward)
 - **Resolution**

Resolution (a simple example)

KB:

(1) father (art, jon)

(2) father (bob, kim)

(3) father (X, Y) \Rightarrow parent (X, Y)

Goal: parent (art, jon)?

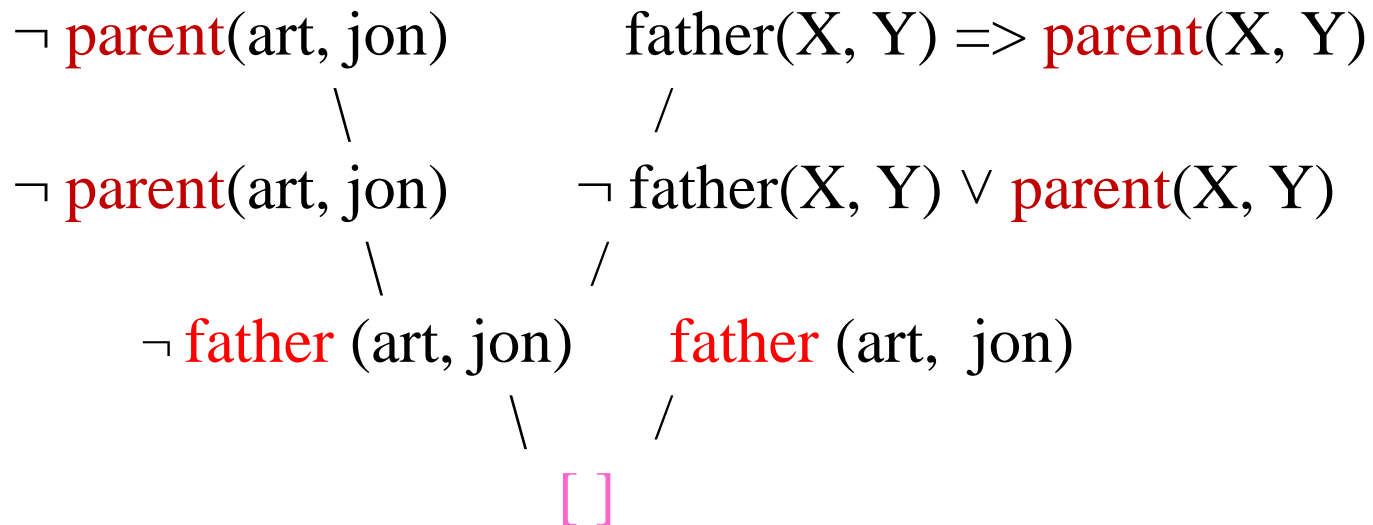
(KB) \wedge (\neg Goal) is "Unsatisfiable"

Resolution (a simple example)

KB:

- (1) father (art, jon)
- (2) father (bob, kim)
- (3) father (X, Y) \Rightarrow parent (X, Y)

Goal: parent (art, jon)?



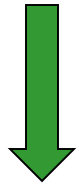
FOL resolution rule

$$\text{UNIFY}(\ell_i, \neg m_j) = \theta.$$

$$\frac{\ell_1 \vee \dots \vee \ell_k, \quad m_1 \vee \dots \vee m_n}{\text{SUBST}(\theta, \ell_1 \vee \dots \vee \ell_{i-1} \vee \ell_{i+1} \vee \dots \vee \ell_k \vee m_1 \vee \dots \vee m_{j-1} \vee m_{j+1} \vee \dots \vee m_n)}$$

- **Example:**

$$[\textit{Animal}(F(x)) \vee \textit{Loves}(G(x), x)] \quad \text{and} \quad [\neg \textit{Loves}(u, v) \vee \neg \textit{Kills}(u, v)]$$



$$\theta = \{u/G(x), v/x\}$$

$$[\textit{Animal}(F(x)) \vee \neg \textit{Kills}(G(x), x)]$$

FOL resolution (example)

$$\frac{\neg Rich(x) \vee Unhappy(x) \quad Rich(Me)}{Unhappy(Me)}$$

with $\theta = \{x/Me\}$

FOL Conjunctive normal form (CNF)

Steps:

- 1.
- 2.
- 3.
- 4.
- 5.
- 6.
- 7.

FOL Conjunctive normal form (CNF)

Steps:

1. Replace $P \Rightarrow Q$ by $\neg P \vee Q$
2. Move \neg inwards, e.g., $\neg \forall x P$ becomes $\exists x \neg P$
3. Standardize variables apart, e.g., $\forall x P \vee \exists x Q$ becomes $\forall x P \vee \exists y Q$
4. Move quantifiers left in order, e.g., $\forall x P \vee \exists x Q$ becomes $\forall x \exists y P \vee Q$
5. Eliminate \exists by Skolemization (next slide)
6. Drop universal quantifiers
7. Distribute \wedge over \vee , e.g., $(P \wedge Q) \vee R$ becomes $(P \vee Q) \wedge (P \vee R)$

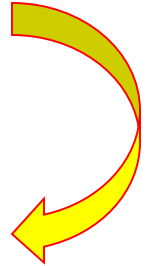
Skolemization

- Why can't (y) be replaced by a constant symbol?

$$\forall x [\exists y \text{ Animal}(y) \wedge \neg \text{Loves}(x, y)] \vee [\exists z \text{ Loves}(z, x)]$$

- Everyone loves the same animal (F), and the same (G) loves everyone

$$\forall x [\text{Animal}(F) \wedge \neg \text{Loves}(x, F)] \vee \text{Loves}(G, x)$$



- Each person loves a different animal $F(x)$, and a different $G(x)$ loves each person

$$\forall x [\text{Animal}(F(x)) \wedge \neg \text{Loves}(x, F(x))] \vee \text{Loves}(G(x), x)$$

- Distribution

$$\forall x [\text{Animal}(F(x)) \vee \text{Loves}(G(x), x)] \wedge [\neg \text{Loves}(x, F(x)) \vee \text{Loves}(G(x), x)]$$

Converting to CNF

$$\forall x [\forall y \text{ Animal}(y) \Rightarrow \text{Loves}(x, y)] \Rightarrow [\exists y \text{ Loves}(y, x)]$$

- Eliminate implications

$$\forall x [\neg \forall y \neg \text{Animal}(y) \vee \text{Loves}(x, y)] \vee [\exists y \text{ Loves}(y, x)]$$

- Move negation inwards

$$\forall x [\exists y \neg(\neg \text{Animal}(y) \vee \text{Loves}(x, y))] \vee [\exists y \text{ Loves}(y, x)] .$$

$$\forall x [\exists y \neg \neg \text{Animal}(y) \wedge \neg \text{Loves}(x, y)] \vee [\exists y \text{ Loves}(y, x)] .$$

$$\forall x [\exists y \text{ Animal}(y) \wedge \neg \text{Loves}(x, y)] \vee [\exists y \text{ Loves}(y, x)] .$$

- Standardize variables

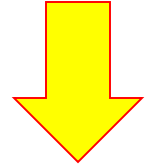
$$\forall x [\exists y \text{ Animal}(y) \wedge \neg \text{Loves}(x, y)] \vee [\exists z \text{ Loves}(z, x)]$$

- Skolemization

$$\forall x [\text{Animal}(F(x)) \wedge \neg \text{Loves}(x, F(x))] \vee \text{Loves}(G(x), x)$$

FOL resolution (another example)

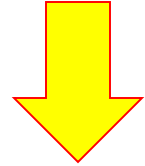
A. $\forall x [\forall y \text{ Animal}(y) \Rightarrow \text{Loves}(x, y)] \Rightarrow [\exists y \text{ Loves}(y, x)]$



Transform to CNF

FOL resolution (another example)

A. $\forall x [\forall y \text{ Animal}(y) \Rightarrow \text{Loves}(x, y)] \Rightarrow [\exists y \text{ Loves}(y, x)]$



Transform to CNF

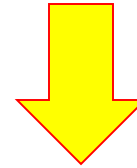
$$\begin{aligned} & \exists y \neg (\text{Animal}(y) \rightarrow \text{Loves}(x, y)) \vee (\exists y \text{ Loves}(y, x)) \\ & \exists y \neg (\text{Animal}(y) \vee \neg \text{Loves}(x, y)) \vee (\exists y \text{ Loves}(y, x)) \\ & \neg (\text{Animal}(\mathbf{F(x)}) \vee \neg \text{Loves}(x, \mathbf{F(x)})) \vee (\text{Loves}(\mathbf{G(x)}, x)) \\ & (\neg \text{Animal}(\mathbf{F(x)}) \wedge \neg \neg \text{Loves}(x, \mathbf{F(x)})) \vee (\text{Loves}(\mathbf{G(x)}, x)) \\ & (\neg \text{Animal}(\mathbf{F(x)}) \vee \text{Loves}(\mathbf{G(x)}, x)) \wedge (\neg \neg \text{Loves}(x, \mathbf{F(x)}) \vee \text{Loves}(\mathbf{G(x)}, x)) \end{aligned}$$

FOL resolution (another example)

A. $\forall x [\forall y \text{ Animal}(y) \Rightarrow \text{Loves}(x, y)] \Rightarrow [\exists y \text{ Loves}(y, x)]$

B. $\forall x [\exists z \text{ Animal}(z) \wedge \text{Kills}(x, z)] \Rightarrow [\forall y \neg \text{Loves}(y, x)]$

C. $\forall x \text{ Animal}(x) \Rightarrow \text{Loves}(\text{Jack}, x)$



Transform to CNF

$$\begin{aligned} & \neg (\exists z \text{ Animal}(z) \wedge \text{Kills}(x, z)) \vee (\forall y \neg \text{Loves}(y, x)) \\ & (\forall z \neg \text{Animal}(z) \vee \neg \text{Kills}(x, z)) \vee (\forall y \neg \text{Loves}(y, x)) \\ & \forall y \forall z (\neg \text{Animal}(z) \vee \neg \text{Kills}(x, z) \vee \neg \text{Loves}(y, x)) \end{aligned}$$

FOL resolution (another example)

- A. $\forall x [\forall y \text{ Animal}(y) \Rightarrow \text{Loves}(x, y)] \Rightarrow [\exists y \text{ Loves}(y, x)]$
- B. $\forall x [\exists z \text{ Animal}(z) \wedge \text{Kills}(x, z)] \Rightarrow [\forall y \neg \text{Loves}(y, x)]$
- C. $\forall x \text{ Animal}(x) \Rightarrow \text{Loves}(\text{Jack}, x)$
- D. $\text{Kills}(\text{Jack}, \text{Tuna}) \vee \text{Kills}(\text{Curiosity}, \text{Tuna})$
- E. $\text{Cat}(\text{Tuna})$
- F. $\forall x \text{ Cat}(x) \Rightarrow \text{Animal}(x)$
- \neg G. $\neg \text{Kills}(\text{Curiosity}, \text{Tuna})$

- A1. $\text{Animal}(F(x)) \vee \text{Loves}(G(x), x)$
- A2. $\neg \text{Loves}(x, F(x)) \vee \text{Loves}(G(x), x)$
- B. $\neg \text{Loves}(y, x) \vee \neg \text{Animal}(z) \vee \neg \text{Kills}(x, z)$
- C. $\neg \text{Animal}(x) \vee \text{Loves}(\text{Jack}, x)$
- D. $\text{Kills}(\text{Jack}, \text{Tuna}) \vee \text{Kills}(\text{Curiosity}, \text{Tuna})$
- E. $\text{Cat}(\text{Tuna})$
- F. $\neg \text{Cat}(x) \vee \text{Animal}(x)$
- \neg G. $\neg \text{Kills}(\text{Curiosity}, \text{Tuna})$

Outline

- What is AI?
- Problem-solving agent
 - Uninformed (DFS), informed (A^*), and local search
 - Adversarial search (minimax, alpha-beta pruning)
- **Knowledge-based agent**
 - Propositional Logic
 - First Order Logic (FOL)
 - **Search Based 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 to use these components to build a *planning agent*?

i.e., an agent that constructs a plan to achieve a goal

Example: Robot Manipulators

- Example: (courtesy of Martin Rohrmeier)



Difference: “search” vs “planning”

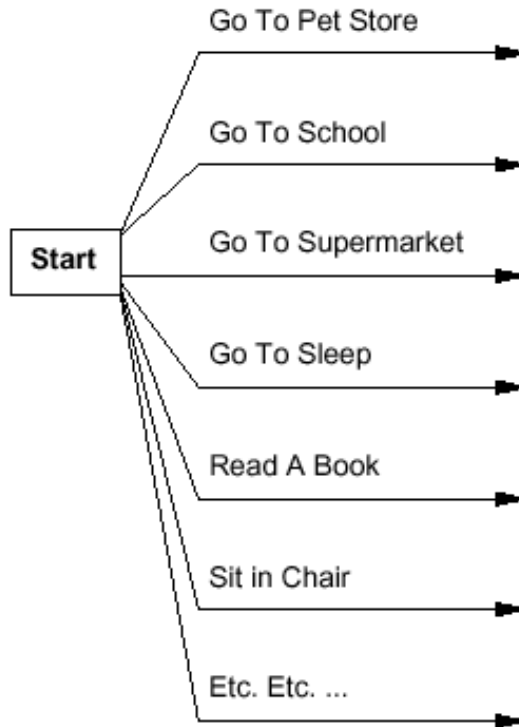
- **Problem-solving agent** can find a sequence of actions that result in a goal state
 - it deals with “**atomic**” representations of states
 - Needs “**domain-specific**” heuristics to perform well in search
- **Planning agent** can also find a sequence of actions that result in a goal state
 - But it uses a “**factored**” representations of states
 - Can have “**generic**” heuristics for search



Logic formulas
in a restricted
format

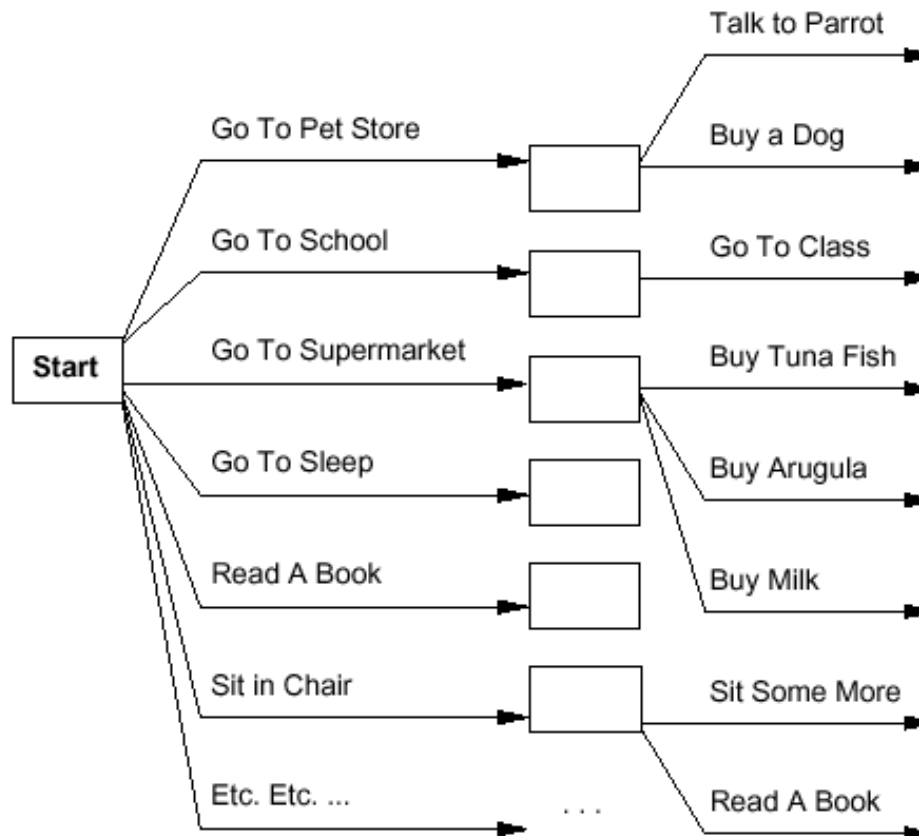
Search vs. planning (example)

- Consider the task **buy milk, bananas, and a cordless drill**, existing search algorithms may fail miserably...



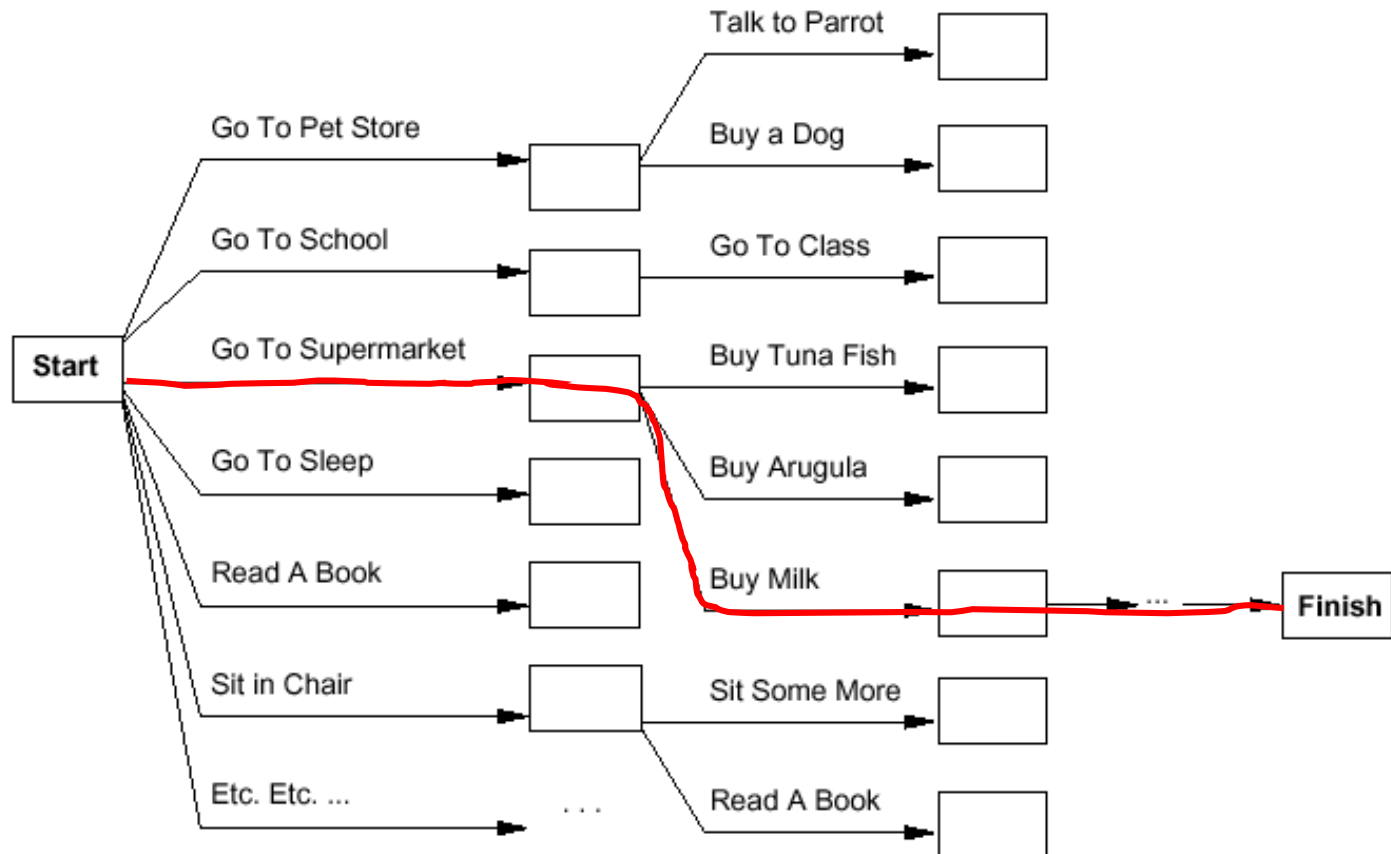
Search vs. planning (example)

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Search vs. planning (example)

- Consider the task **buy milk, bananas, and a cordless drill**, existing search algorithms may fail miserably...



Search vs. planning

- Planning opens up **action** and **goal** representations

	Search	Planning
States		
Actions		
Goal		
Plan		

Search vs. planning

- Planning opens up **action** and **goal** representations

	Search	Planning
States	data structures	Logical sentences
Actions	code	Preconditions/outcomes
Goal	code	Logical sentence (conjunction)
Plan	Sequence from S_0	Constraints on actions

PDDL: *Planning Domain Definition Language*

- It uses a restricted subset of first-order logic (FOL) to make planning **efficiently solvable**

- **State:** a conjunction of functionless ground literals

$Poor \wedge Unknown$

$At(Truck_1, Melbourne) \wedge At(Truck_2, Sydney)$

$At(x, y)$  cannot have variables (x,y)

$At(Father(Fred), Sydney)$  cannot have function symbol

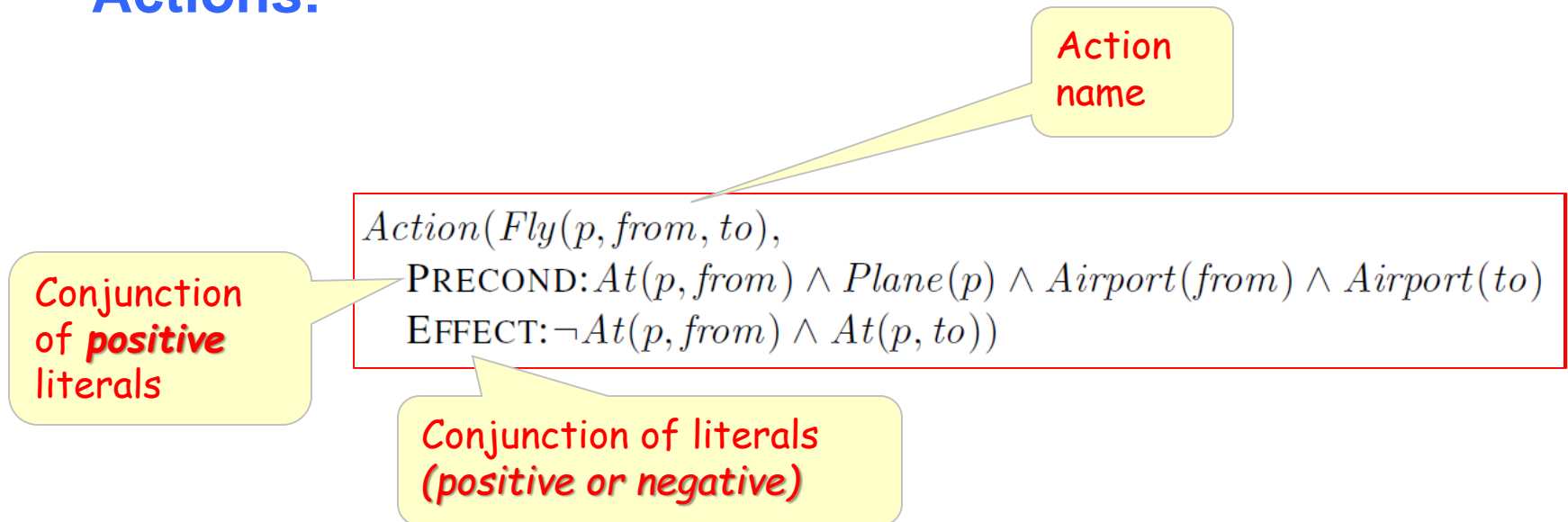
- **Goal:** a conjunction of literals, but may have variables

$At(Home) \wedge Have(Milk) \wedge Have(Bananas) \wedge Have(Drill)$

$At(x) \wedge Sells(x, Milk)$

PDDL: *Planning Domain Definition Language*

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- **State:** a conjunction of functionless ground literals
- **Actions:**



PDDL: *Planning Domain Definition Language*

- It uses a restricted subset of first-order logic (FOL) to make planning **efficiently solvable**
- **State:** a conjunction of functionless ground literals
- **Actions:**

Action(Fly(p, from, to),
PRECOND: $At(p, from) \wedge Plane(p) \wedge Airport(from) \wedge Airport(to)$
EFFECT: $\neg At(p, from) \wedge At(p, to)$

Negative literal

DEL this literal
from the new state

Negative literal

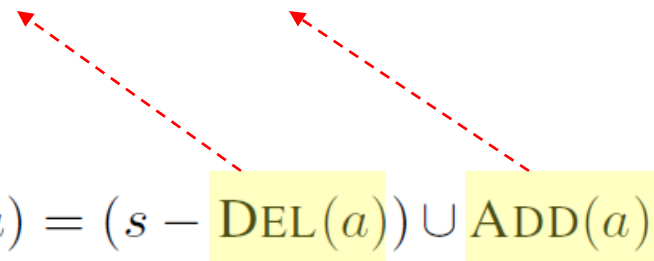
ADD this literal
into the new state

PDDL: *Planning Domain Definition Language*

- It uses a restricted subset of first-order logic (FOL) to make planning **efficiently solvable**
- **State:** a conjunction of functionless ground literals
- **Actions:**

$Action(Fly(p, from, to),$
PRECOND: $At(p, from) \wedge Plane(p) \wedge Airport(from) \wedge Airport(to)$
EFFECT: $\neg At(p, from) \wedge At(p, to)$

- **Transition model:**

$$RESULT(s, a) = (s - \text{DEL}(a)) \cup \text{ADD}(a)$$


Example: *Air cargo transportation planning*



Example: *Air cargo transportation planning*

$Init(At(C_1, SFO) \wedge At(C_2, JFK) \wedge At(P_1, SFO) \wedge At(P_2, JFK) \\ \wedge Cargo(C_1) \wedge Cargo(C_2) \wedge Plane(P_1) \wedge Plane(P_2) \\ \wedge Airport(JFK) \wedge Airport(SFO))$

$Goal(At(C_1, JFK) \wedge At(C_2, SFO))$

$Action(Load(c, p, a),$

PRECOND: $At(c, a) \wedge At(p, a) \wedge Cargo(c) \wedge Plane(p) \wedge Airport(a)$

EFFECT: $\neg At(c, a) \wedge In(c, p)$)

$Action(Unload(c, p, a),$

PRECOND: $In(c, p) \wedge At(p, a) \wedge Cargo(c) \wedge Plane(p) \wedge Airport(a)$

EFFECT: $At(c, a) \wedge \neg In(c, p)$)

$Action(Fly(p, from, to),$

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Example: *Air cargo transportation planning*

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 $\wedge Cargo(C_1) \wedge Cargo(C_2) \wedge Plane(P_1) \wedge Plane(P_2)$
 $\wedge Airport(JFK) \wedge Airport(SFO))$

$Goal(At(C_1, JFK) \wedge At(C_2, SFO))$

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Example: *Air cargo transportation planning*

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 $\wedge Cargo(C_1) \wedge Cargo(C_2) \wedge Plane(P_1) \wedge Plane(P_2)$
 $\wedge Airport(JFK) \wedge Airport(SFO))$

$Goal(At(C_1, JFK) \wedge At(C_2, SFO))$

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Example: *Air cargo transportation planning*

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 $\wedge Airport(JFK) \wedge Airport(SFO))$

$Goal(At(C_1, JFK) \wedge At(C_2, SFO))$

$Action(Load(c, p, a),$

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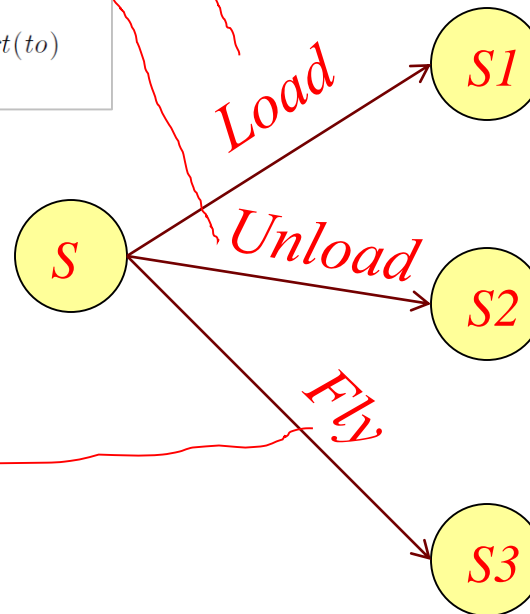
$Action(Fly(p, from, to),$

PRECOND: $At(p, from) \wedge Plane(p) \wedge Airport(from) \wedge Airport(to)$

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Example: Air cargo transportation planning

$Init(At(C_1, SFO) \wedge At(C_2, JFK) \wedge At(P_1, SFO) \wedge At(P_2, JFK)$
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 $Goal(At(C_1, JFK) \wedge At(C_2, SFO))$
 $Action(Load(c, p, a),$
 PRECOND: $At(c, a) \wedge At(p, a) \wedge Cargo(c) \wedge Plane(p) \wedge Airport(a)$
 EFFECT: $\neg At(c, a) \wedge In(c, p)$)
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 EFFECT: $At(c, a) \wedge \neg In(c, p)$)
 $Action(Fly(p, from, to),$
 PRECOND: $At(p, from) \wedge Plane(p) \wedge Airport(from) \wedge Airport(to)$
 EFFECT: $\neg At(p, from) \wedge At(p, to)$)



The following plan is a solution to the problem:

$[Load(C_1, P_1, SFO), Fly(P_1, SFO, JFK), Unload(C_1, P_1, JFK),$
 $Load(C_2, P_2, JFK), Fly(P_2, JFK, SFO), Unload(C_2, P_2, SFO)]$.

Example: *Changing the spare tire*



Example: *Changing the spare tire*

Init(Tire(Flat) \wedge Tire(Spare) \wedge At(Flat, Axle) \wedge At(Spare, Trunk))

Goal(At(Spare, Axle))

Action(Remove(obj, loc),

 PRECOND: *At(obj, loc)*

 EFFECT: \neg *At(obj, loc)* \wedge *At(obj, Ground)*)

Action(PutOn(t, Axle),

 PRECOND: *Tire(t)* \wedge *At(t, Ground)* \wedge \neg *At(Flat, Axle)*

 EFFECT: \neg *At(t, Ground)* \wedge *At(t, Axle)*)

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)* \wedge \neg *At(Flat, Trunk)*)

Example: *Changing the spare tire*

Init(*Tire*(*Flat*) \wedge *Tire*(*Spare*) \wedge *At*(*Flat*, *Axle*) \wedge *At*(*Spare*, *Trunk*))

Goal(*At*(*Spare*, *Axle*))

Action(*Remove*(*obj*, *loc*),

PRECOND: *At*(*obj*, *loc*)

EFFECT: \neg *At*(*obj*, *loc*) \wedge *At*(*obj*, *Ground*))

Action(*PutOn*(*t*, *Axle*),

PRECOND: *Tire*(*t*) \wedge *At*(*t*, *Ground*) \wedge \neg *At*(*Flat*, *Axle*)

EFFECT: \neg *At*(*t*, *Ground*) \wedge *At*(*t*, *Axle*))

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*) \wedge \neg *At*(*Flat*, *Trunk*))

Example: Changing the spare tire

$Init(Tire(Flat) \wedge Tire(Spare) \wedge At(Flat, Axle) \wedge At(Spare, Trunk))$
 $Goal(At(Spare, Axle))$

$Action(Remove(obj, loc),$
 PRECOND: $At(obj, loc)$
 EFFECT: $\neg At(obj, loc) \wedge At(obj, Ground))$

$Action(PutOn(t, Axle),$
 PRECOND: $Tire(t) \wedge At(t, Ground) \wedge \neg At(Flat, Axle)$
 EFFECT: $\neg At(t, Ground) \wedge At(t, Axle)$

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

Example: Changing the spare tire

Init($Tire(Flat) \wedge Tire(Spare) \wedge At(Flat, Axle) \wedge At(Spare, Trunk)$)

Goal($At(Spare, Axle)$)

Action(*Remove*(*obj*, *loc*),

PRECOND: $At(obj, loc)$

EFFECT: $\neg At(obj, loc) \wedge At(obj, Ground)$)

Action(*PutOn*(*t*, *Axle*),

PRECOND: $Tire(t) \wedge At(t, Ground) \wedge \neg At(Flat, Axle)$

EFFECT: $\neg At(t, Ground) \wedge At(t, Axle)$)

Action(*LeaveOvernight*,

PRECOND:

EFFECT: $\neg At(Spare, Ground) \wedge \neg At(Spare, Axle) \wedge \neg At(Spare, Trunk)$
 $\wedge \neg At(Flat, Ground) \wedge \neg At(Flat, Axle) \wedge \neg At(Flat, Trunk)$)

Example: Changing the spare tire

Init($Tire(Flat) \wedge Tire(Spare) \wedge At(Flat, Axle) \wedge At(Spare, Trunk)$)

Goal($At(Spare, Axle)$)

Action(*Remove*(*obj*, *loc*),

 PRECOND: $At(obj, loc)$

 EFFECT: $\neg At(obj, loc) \wedge At(obj, Ground)$)

Action(*PutOn*(*t*, *Axle*),

 PRECOND: $Tire(t) \wedge At(t, Ground) \wedge \neg At(Flat, Axle)$

 EFFECT: $\neg At(t, Ground) \wedge At(t, Axle)$)

Action(*LeaveOvernight*,

 PRECOND:

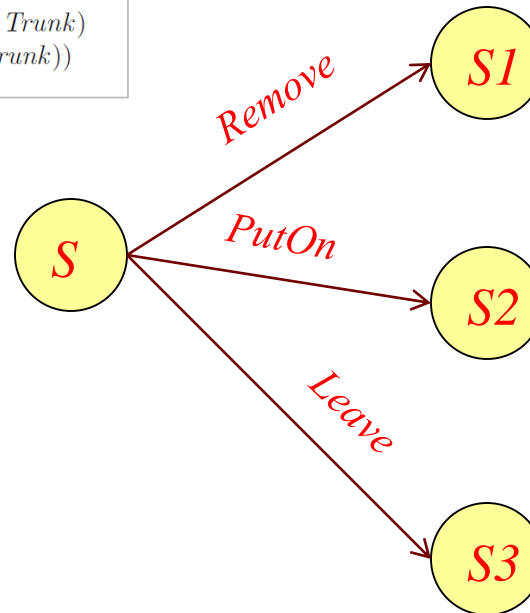
 EFFECT: $\neg At(Spare, Ground) \wedge \neg At(Spare, Axle) \wedge \neg At(Spare, Trunk)$

$\wedge \neg At(Flat, Ground) \wedge \neg At(Flat, Axle) \wedge \neg At(Flat, Trunk)$)

Bad neighborhood

Example: Changing the spare tire

```
Init(Tire(Flat)  $\wedge$  Tire(Spare)  $\wedge$  At(Flat, Axle)  $\wedge$  At(Spare, Trunk))  
Goal(At(Spare, Axle))  
Action(Remove(obj, loc),  
  PRECOND: At(obj, loc)  
  EFFECT:  $\neg$  At(obj, loc)  $\wedge$  At(obj, Ground))  
Action(PutOn(t, Axle),  
  PRECOND: Tire(t)  $\wedge$  At(t, Ground)  $\wedge$   $\neg$  At(Flat, Axle)  
  EFFECT:  $\neg$  At(t, Ground)  $\wedge$  At(t, Axle))  
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)  $\wedge$   $\neg$  At(Flat, Trunk))
```

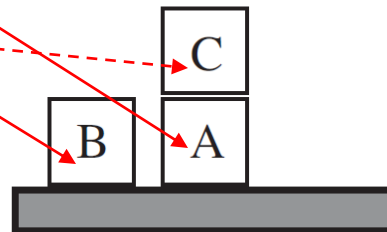


A solution to the problem is $[Remove(Flat, Axle), Remove(Spare, Trunk), PutOn(Spare, Axle)]$.

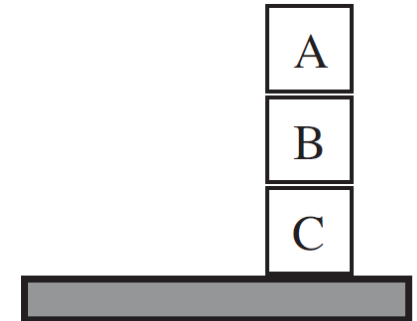
Example: *the blocks world*

Relations

- $\text{On}(A, \text{Table})$
- $\text{On}(B, \text{Table})$
- $\text{On}(C, A)$
- $\text{Clear}(B)$
- $\text{Clear}(C)$
- $\text{Block}(A), \text{Block}(B), \text{Block}(C)$

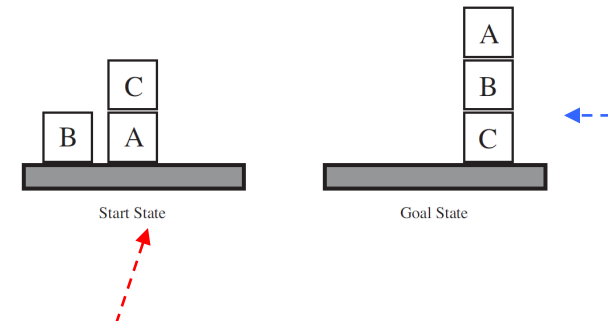


Start State



Goal State

Example: *the blocks world*



$Init(On(A, Table) \wedge On(B, Table) \wedge On(C, A)$
 $\wedge Block(A) \wedge Block(B) \wedge Block(C) \wedge Clear(B) \wedge Clear(C))$

$Goal(On(A, B) \wedge On(B, C))$

$Action(Move(b, x, y),$

PRECOND: $On(b, x) \wedge Clear(b) \wedge Clear(y) \wedge Block(b) \wedge Block(y) \wedge$
 $(b \neq x) \wedge (b \neq y) \wedge (x \neq y),$

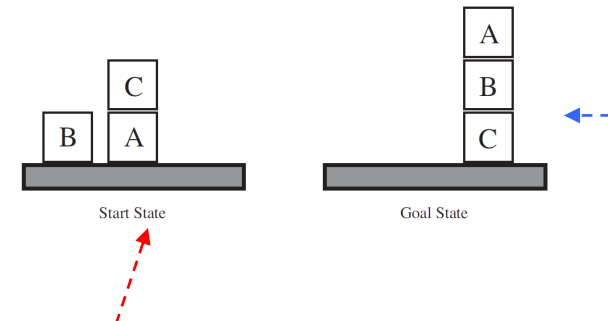
EFFECT: $On(b, y) \wedge Clear(x) \wedge \neg On(b, x) \wedge \neg Clear(y))$

$Action(MoveToTable(b, x),$

PRECOND: $On(b, x) \wedge Clear(b) \wedge Block(b) \wedge (b \neq x),$

EFFECT: $On(b, Table) \wedge Clear(x) \wedge \neg On(b, x))$

Example: *the blocks world*



$Init(On(A, Table) \wedge On(B, Table) \wedge On(C, A)$
 $\wedge Block(A) \wedge Block(B) \wedge Block(C) \wedge Clear(B) \wedge Clear(C))$

$Goal(On(A, B) \wedge On(B, C))$

$Action(Move(b, x, y),$

PRECOND: $On(b, x) \wedge Clear(b) \wedge Clear(y) \wedge Block(b) \wedge Block(y) \wedge$
 $(b \neq x) \wedge (b \neq y) \wedge (x \neq y),$

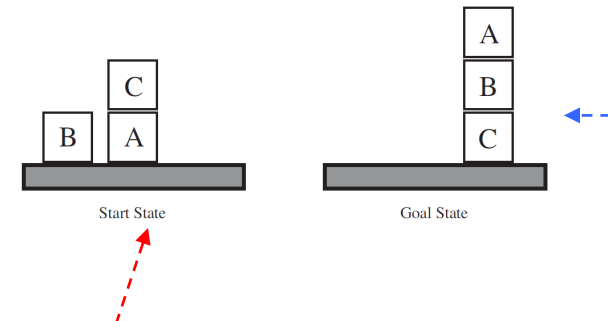
EFFECT: $On(b, y) \wedge Clear(x) \wedge \neg On(b, x) \wedge \neg Clear(y))$

$Action(MoveToTable(b, x),$

PRECOND: $On(b, x) \wedge Clear(b) \wedge Block(b) \wedge (b \neq x),$

EFFECT: $On(b, Table) \wedge Clear(x) \wedge \neg On(b, x))$

Example: *the blocks world*



$Init(On(A, Table) \wedge On(B, Table) \wedge On(C, A)$
 $\wedge Block(A) \wedge Block(B) \wedge Block(C) \wedge Clear(B) \wedge Clear(C))$

$Goal(On(A, B) \wedge On(B, C))$

$Action(Move(b, x, y),$

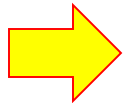
PRECOND: $On(b, x) \wedge Clear(b) \wedge Clear(y) \wedge Block(b) \wedge Block(y) \wedge$
 $(b \neq x) \wedge (b \neq y) \wedge (x \neq y),$

EFFECT: $On(b, y) \wedge Clear(x) \wedge \neg On(b, x) \wedge \neg Clear(y))$

$Action(MoveToTable(b, x),$

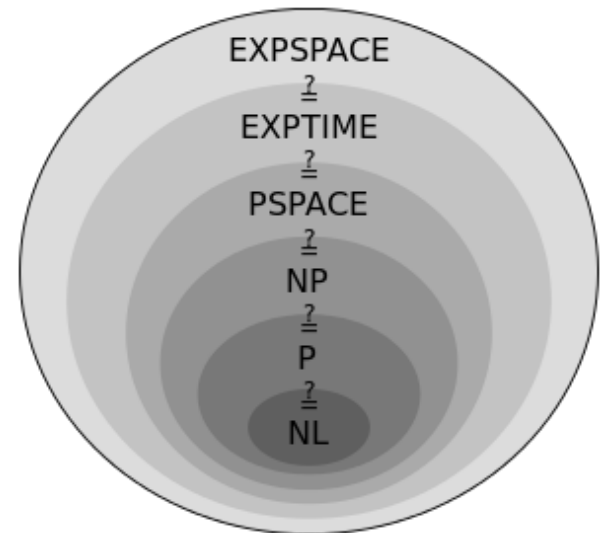
PRECOND: $On(b, x) \wedge Clear(b) \wedge Block(b) \wedge (b \neq x),$

EFFECT: $On(b, Table) \wedge Clear(x) \wedge \neg On(b, x))$

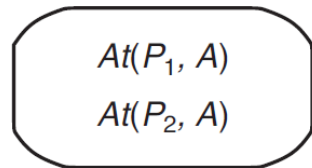


Complexity of classic planning

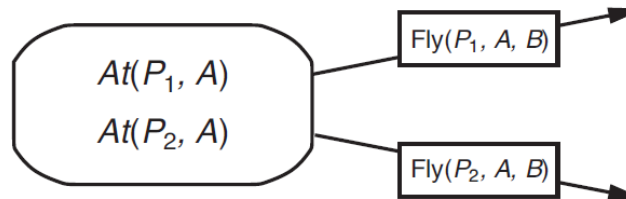
- PSPACE, a complexity class that is larger/harder than NP
 - **Planner**: ask for a sequence of actions that, if executed from a state, will make goal become true in a future state
 - PSPACE
 - **Theorem prover**: ask if a sentence is true given KB (does not have the notion of state transition)
 - NP



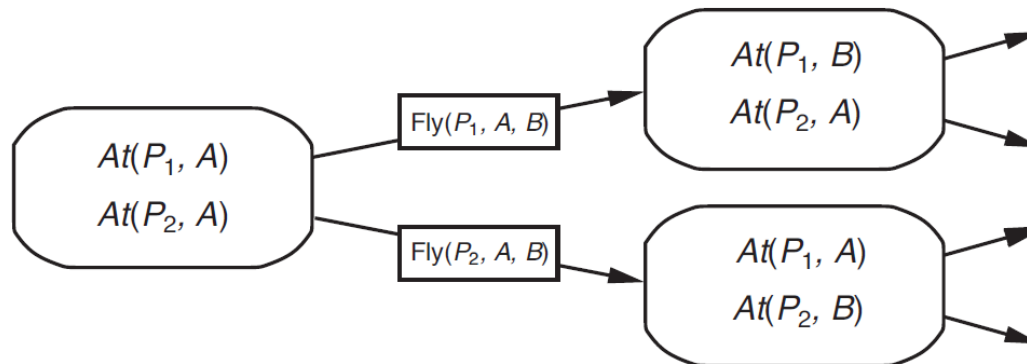
Planning as state-space search (*forward*)



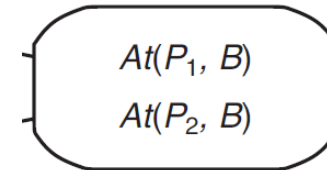
Planning as state-space search (*forward*)



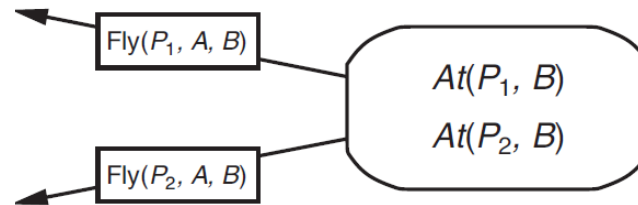
Planning as state-space search (*forward*)



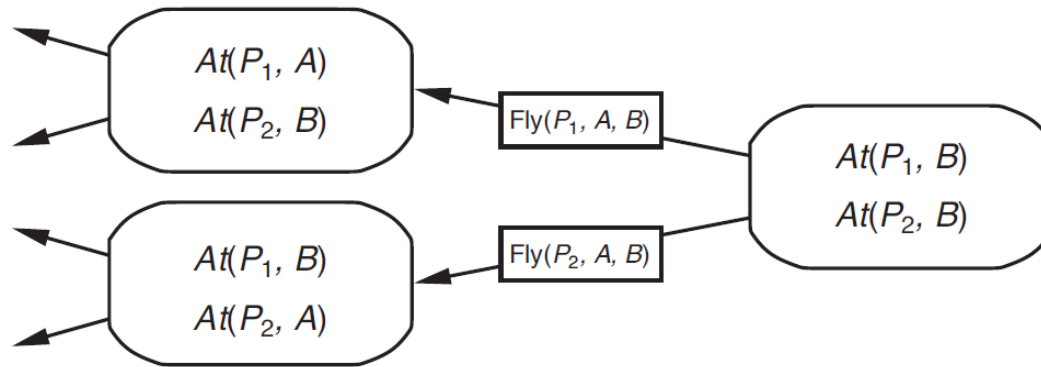
Planning as state-space search (*backward*)



Planning as state-space search (*backward*)



Planning as state-space search (*backward*)



Heuristics for planning

- Neither forward nor backward search is efficient without a good heuristic function
 - Need an admissible heuristic
 - i.e., never overestimate the distance from a state (s) to the goal

Planning graph

- It is a **data structure** used to give **heuristic estimates**
 - Can be applied to any of the search techniques
 - Will never overestimate; and often very accurate

```
Init(Have(Cake))  
Goal(Have(Cake) ∧ Eaten(Cake))  
Action(Eat(Cake))  
  PRECOND: Have(Cake)  
  EFFECT:  $\neg Have(Cake) \wedge Eaten(Cake)$   
Action(Bake(Cake))  
  PRECOND:  $\neg Have(Cake)$   
  EFFECT: Have(Cake)
```

Planning graph

- S_0, S_1, S_2 – states
 - May be reachable at each level
 - Mutual exclusion (mutex) links
- A_0, A_1 – actions
 - Mutual exclusion (mutex) links

```
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:  $\neg$  Have(Cake)
  EFFECT: Have(Cake)
```

S_0

A_0

S_1

A_1

S_2

$Have(Cake)$

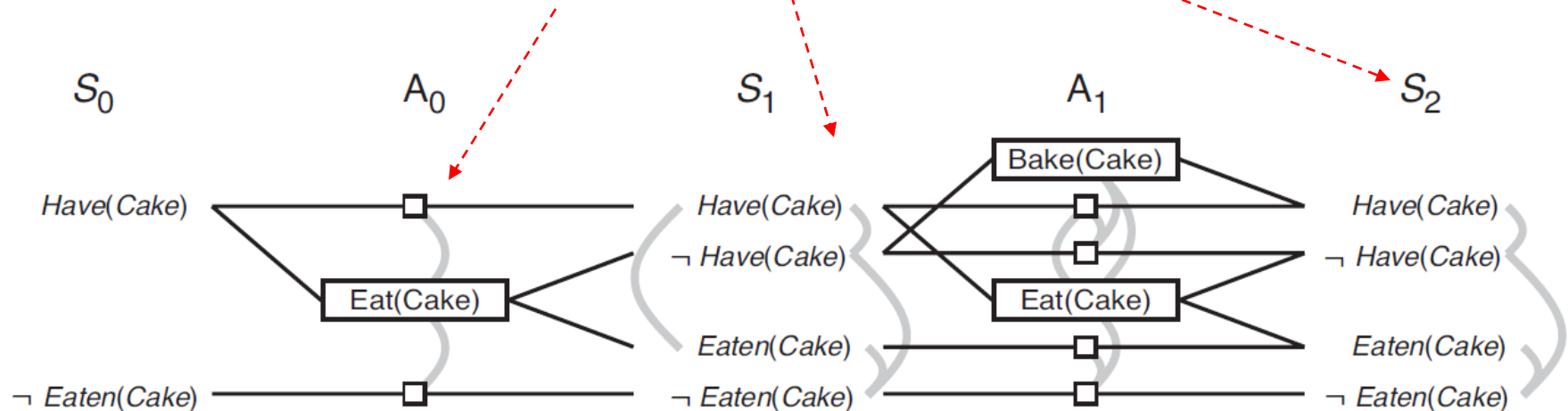
$\neg Eaten(Cake)$

Planning graph

- S_0, S_1, S_2 – states
 - May be reachable at each level
 - Mutual exclusion (mutex) links
- A_0, A_1 – actions
 - Mutual exclusion (mutex) links

```

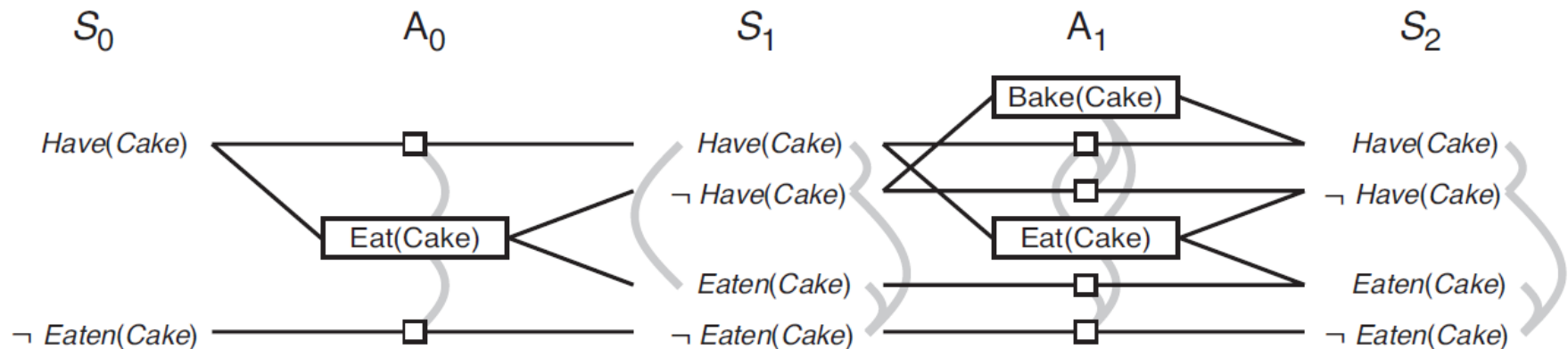
Init(Have(Cake))
Goal(Have(Cake)  $\wedge$  Eaten(Cake))
Action(Eat(Cake))
  PRECOND: Have(Cake)
  EFFECT:  $\neg$  Have(Cake)  $\wedge$  Eaten(Cake)
Action(Bake(Cake))
  PRECOND:  $\neg$  Have(Cake)
  EFFECT: Have(Cake)
    
```



Planning graph: *mutex actions*

- Effects contradict each other
 - $\text{Eat}(\text{Cake}).\text{effect}$ vs. $\text{Have}(\text{Cake}).\text{effect}$
- Preconditions contradict each other
 - $\text{Bake}(\text{Cake}).\text{precond}$ vs $\text{Eat}(\text{Cake}).\text{precond}$
- Interference (one action's effect contradicts the other action's precondition)
 - $\text{Eat}(\text{Cake}).\text{effect}$ vs $\text{Have}(\text{Cake}).\text{precond}$

```
Init(Have(Cake))
Goal(Have(Cake) ∧ Eaten(Cake))
Action(Eat(Cake))
  PRECOND: Have(Cake)
  EFFECT: ¬ Have(Cake) ∧ Eaten(Cake)
Action(Bake(Cake))
  PRECOND: ¬ Have(Cake)
  EFFECT: Have(Cake)
```



Properties of a planning graph

- **Polynomial** in the size of the planning problem
 - Instead of being “**exponential**” in size
- If any goal literal fails to appear in the final level of the graph, then the problem is **unsolvable**
- The **cost** of achieving any **goal (g)** can be **estimated** as the level at which (g) first appears in the planning graph constructed from (s) as the initial state

Graph planning algorithm

function GRAPHPLAN(*problem*) **returns** solution or failure

graph \leftarrow INITIAL-PLANNING-GRAPH(*problem*)

goals \leftarrow CONJUNCTS(*problem*.GOAL)

nogoods \leftarrow an empty hash table

for *tl* = 0 **to** ∞ **do**

if *goals* all non-mutex in S_t of *graph* **then**

solution \leftarrow EXTRACT-SOLUTION(*graph*, *goals*, NUMLEVELS(*graph*), *nogoods*)

if *solution* \neq failure **then return** *solution*

if *graph* and *nogoods* have both leveled off **then return** failure

graph \leftarrow EXPAND-GRAPH(*graph*, *problem*)

Example planning graph (spare tire)

S_0

$At(Spare, Trunk).$

A_0

S_1

A_1

S_2

$At(Flat, Axle).$

$\neg At(Spare, Axle)$

$\neg At(Flat, Ground)$

$\neg At(Spare, Ground)$

Example planning graph (spare tire)

