

# **CS 4649/7649 RIP**

## **Robot Intelligence: Planning**

### **Partial Order Planning, Graphplan**

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S. Joo (sungmoon.joo@cc.gatech.edu)  
\*Slides based on Dr. Mike Stilman's lecture slides

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1

### **Course Info.**

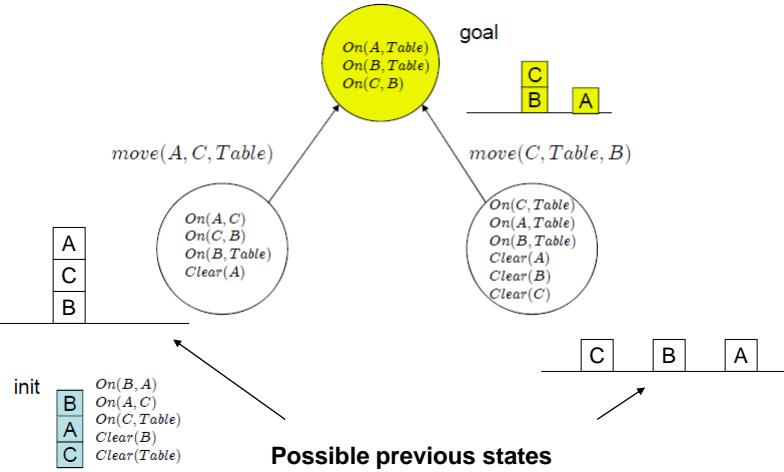
- **Course Website:** [joosm.github.io/RIP2014](https://joosm.github.io/RIP2014)
- **Course Wiki:** [github.com/RIP2014/RIP2014Wiki/wiki](https://github.com/RIP2014/RIP2014Wiki/wiki)
  - add your contact info, start grouping/filling in project ideas, etc.
  - github invitation sent (if you didn't get one, let me know)
  - \*announcements on t-square
- **Email me**([sungmoon.joo@cc.gatech.edu](mailto:sungmoon.joo@cc.gatech.edu))
  - Introduce yourself, your github id
  - Project ideas, etc.

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2

## Regression Planning

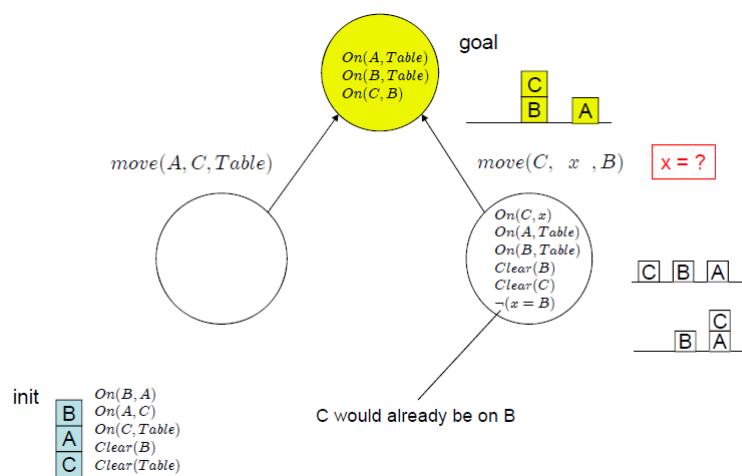


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3

## Regression Planning (Variables Show Up)



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4

## Regression Planning

- Just as with forward (progression) planning this algorithm would be complete. However:
- Still large branching factor (potentially larger)
- When do we instantiate the variables?

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5

## Regression Planning

- Still large branching factor (potentially larger)
- When do we instantiate the variables?
- Introduces new concept:

### **"Least Commitment Planning"**

Make choices **only** when they are relevant to solving  
The current part of the problem

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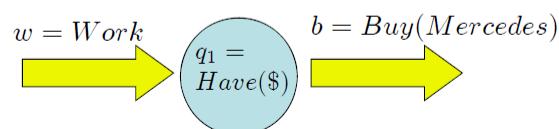
6

## Partial Order Planning

- Based on the concept of “Least Commitment”
- Nodes are partial plans
- Arcs/Transitions are **plan refinements**
- Solution is a node, not a path (search in plan space!!)

## Partial Order Planning

- A plan consists of:
  - **A: Set of actions**
  - **O: Set of orderings for actions ( $a < b$ )**
  - **Q: Set of causal links**
- A causal link  $q_1 \in Q$  is defined as follows:
  - Action b has a precondition that is established by Action w



## Partial Order Planning: Threats

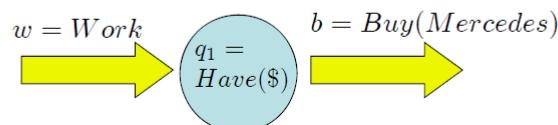
- A threat to  $(a, q, b)$  is defined as an action  $t$  that:
  - Has  $\neg q$  as an effect

$$\neg q \in t_{\text{add}}$$

- Could occur between  $a$  and  $b$

$O \cup a < t < b$  is inconsistent

- What action  $t$  would be a threat to causal link  $q_1$ ?



## Partial Order Planning: Initialization

Since we're only talking about actions, let's turn states into actions:

- $A_0$ 
  - **No preconditions**
  - Initial state as **effects**
  - Must be the **first step** in the plan (all actions  $> A_0$  in  $O$ )
- $A_N$ 
  - **No effects**
  - Goals as **preconditions**
  - Must be the **last step** in the plan (all actions  $< A_N$  in  $O$ )

## POP Algorithm: Simplified UCPOP<sup>\*</sup> (Weld)

```

POP ((A,O,Q), agenda, actions)
If agenda=∅ then return(A,O,Q)
(q,aneed) = Choose(pair) from agenda
aadd = Choose(actions) s.t. q ∈ Add(aadd)
If no such action aadd exists, Fail

Q' = Q ∪ (aadd,q,aneed)
O' = O ∪ (aadd < aneed)
agenda' = agenda - (q,aneed)

If aadd is new, then A = A ∪ aadd and
    ∀ p ∈ PC(aadd), add (p, aadd) to agenda'

For every action ai that threatens any causal link (ai, Q, aj) in Q'
    Choose to add ai < aj or ai < at to O
        if neither choice is consistent, Fail

POP((A',O',Q'), agenda, actions)

```

The magic “Choose” enables backtracking

\*UCPOP (Universal, Conditional Partial-Order Planner)

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11

## POP Algorithm

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        if neither choice is consistent, Fail
    buy(x) → PC: Have ($)
    D: Have ($)
    A: Have(x) buy(Mercedes) → aN = Finish
    PC : Have(Mercedes) →

POP((A',O',Q'), agenda, actions)

```

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12

## POP Algorithm

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```

If a<sub>add</sub> is new, then A = A ∪ a<sub>add</sub> and  
 $\forall p \in PC(a_{add})$ , add (p, a<sub>add</sub>) to agenda'

For every action a, that threatens any causal link (a<sub>i</sub>, Q, a<sub>j</sub>) in Q'  
 Choose to add a<sub>i</sub> < a<sub>j</sub> or a<sub>j</sub> < a<sub>i</sub> to O  
 if neither choice is consistent, Fail

POP((A',O',Q'), agenda, actions)

work →  
A : Have(\$)

Agenda = {  
}

Q = {(buy(Merc.), Have(Merc.), Finish)  
 (work, Have(\$), buy(Merc.))}  
{}

O = {  
 buy(Merc.) < Finish  
 work < buy(Merc.)  
{}}

*a<sub>N</sub> = Finish*  
*PC : Have(Mercedes)*

buy(Mercedes) →      work →

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13

## POP Algorithm

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 if neither choice is consistent, Fail

POP((A',O',Q'), agenda, actions)

Agenda = {  
}

Q = {(buy(Merc.), Have(Merc.), Finish)  
 (work, Have(\$), buy(Merc.))}  
{}

O = {  
 buy(Merc.) < Finish  
 work < buy(Merc.)  
{}}

work →  
**DONE!** →      buy(Mercedes) →      *a<sub>N</sub> = Finish*  
*PC : Have(Mercedes)*

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14

## POP Algorithm Details

Protect Causal Links  
use O:

Demotion:

$$a_t < a_i$$

Promotion:

$$a_t > a_j$$

```

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  POP((A',O',Q'), agenda, actions)

```

POP terminates

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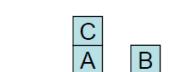
15

## POP solves Sussman Anomaly

$A_0$

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

Init



Goal



$On(A, B)$     $On(B, C)$

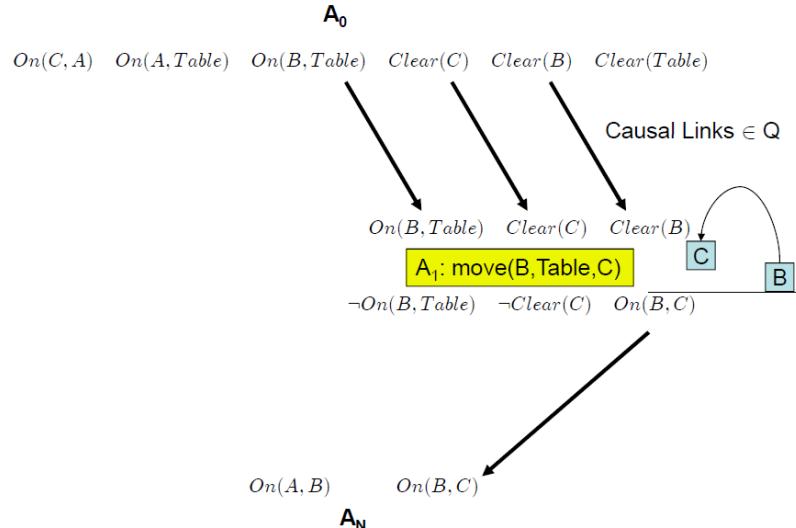
$A_N$

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16

## POP solves Sussman Anomaly

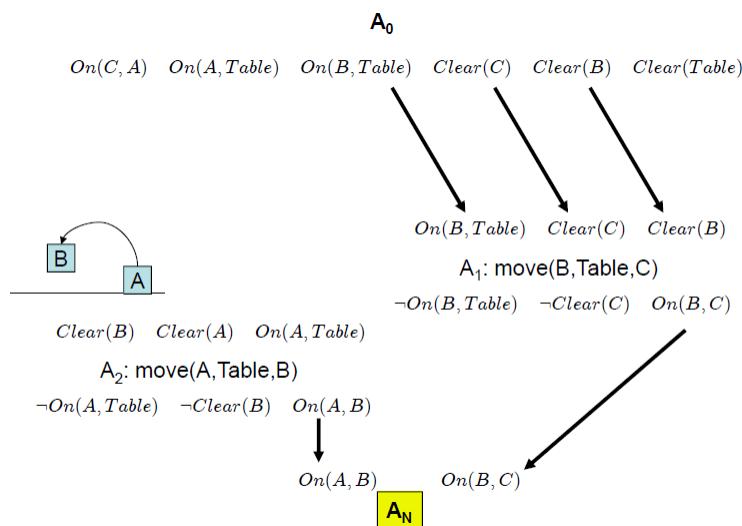


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17

## POP solves Sussman Anomaly



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18

## POP Algorithm Details

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    if neither choice is consistent, Fail

  POP((A',O',Q'), agenda, actions)

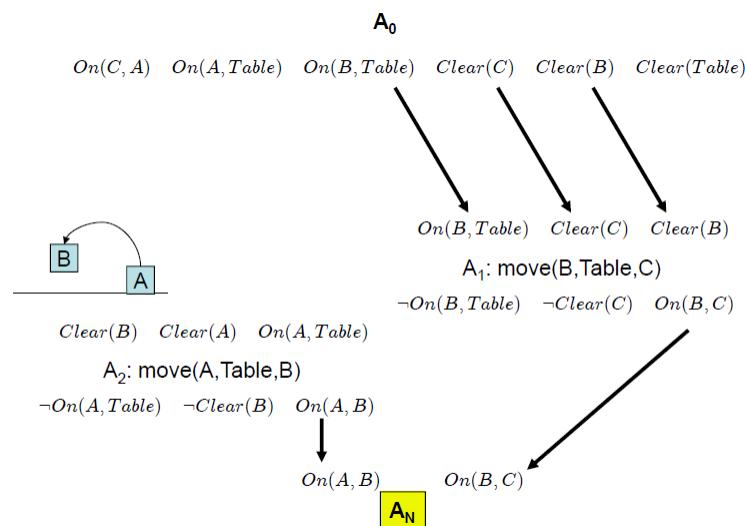
```

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19

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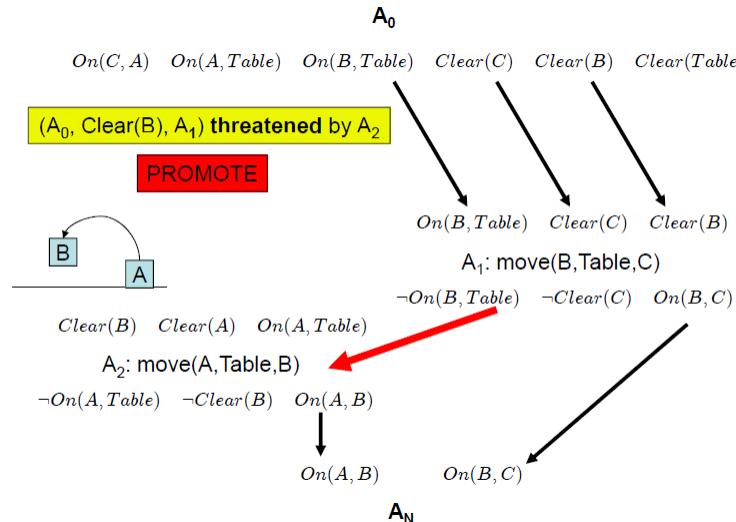


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20

## POP solves Sussman Anomaly

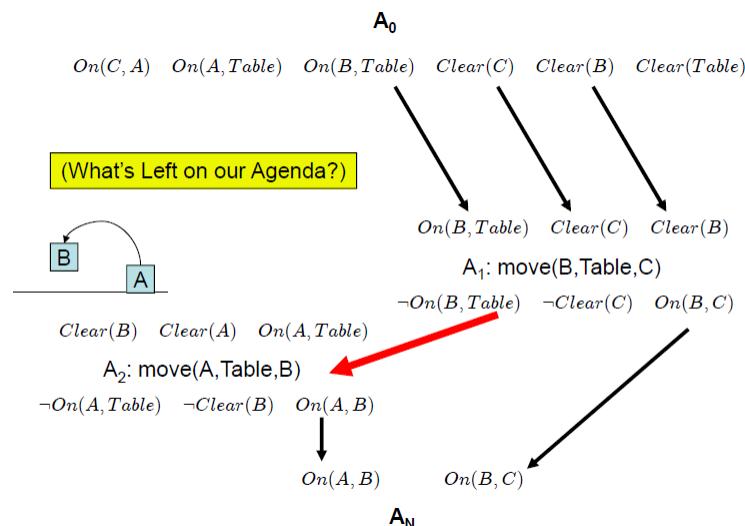


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21

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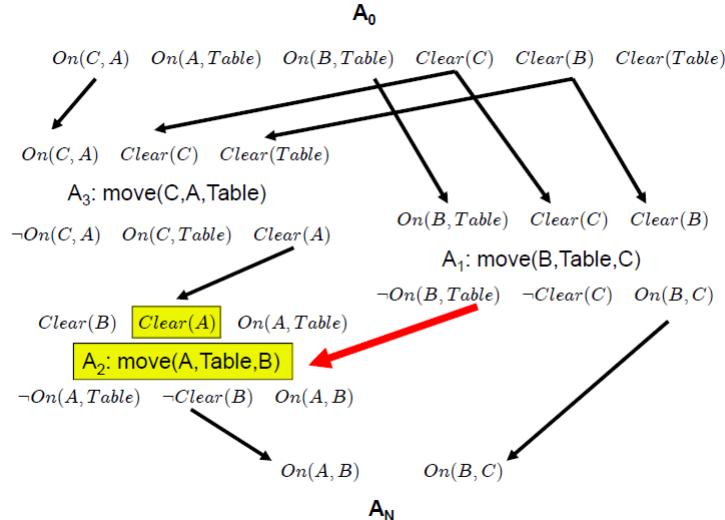


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22

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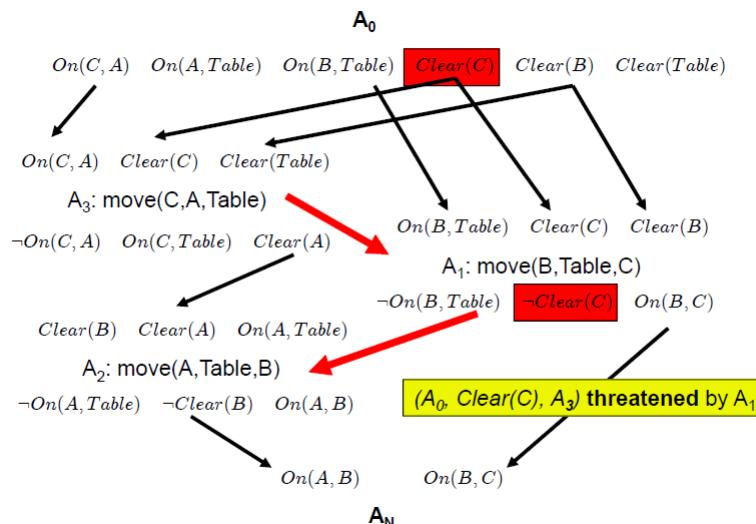


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23

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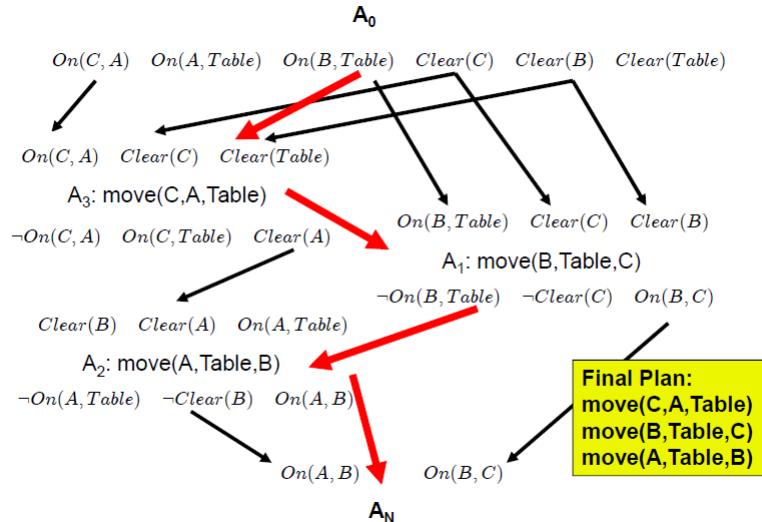


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24

## POP solves Sussman Anomaly



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25

## Properties of a Planner

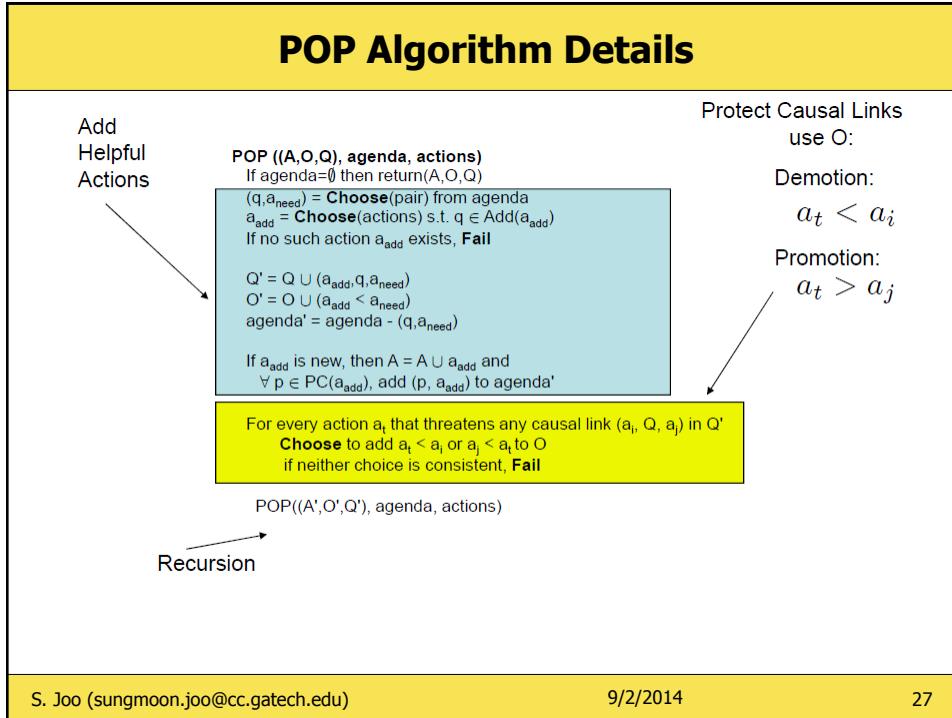
- 1) Sound: The planner produces valid plans
- 2) Optimal: The planner produces optimal (shortest) plans
- 3) Complete: The planner finds a solution when there is one or returns that the solution is not possible.

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26

## POP Algorithm Details



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27

## Summary

- Important Properties of Planners
  - Soundness
  - Completeness
  - Optimality
- Two types of planning:
  - State Space
    - Non-deterministic Choices:  $n = |\text{actions}|$
    - Backtracking for goal ordering
    - Simplicity!
  - Plan Space
    - Least Commitment
    - Non-deterministic Choices:  $n = |\text{preconditions}| + |\text{link protection}|$
    - Smaller branching factor (goal ordering not relevant)
    - Typically more optimal

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28

## State Space vs. Plan Space

	State Space	Plan Space
Algorithm	Progression Regression	Partial Order Planning
Nodes	World States	Partial Plans
Transitions	Actions -Move(x,y,z) -Load(x,y) -Open(r)	Plan Refinement Ops -Adding Steps -Promotion -Demotion

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29

## Last Major Classical Planner: Graphplan\*

- A form of state space planning: actually "Factored State Space"
- Nodes are **literals** and **instantiated actions**
- Uses Least Commitment principle to constrain search
- Simpler than Plan Space Planning
- More general than State Space Planning



Merrick Furst



Avrim Blum

1997

<http://www.cs.cmu.edu/~avrim/graphplan.html>

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30

## Graphplan Basics

- Graph has levels that represent time steps
- Levels alternate between preconditions and actions  
 $P_0 \ A_0 \ P_1 \ A_1 \ P_2 \ A_2 \dots$
- Precondition level N reflects what **could** be true in N steps
- Mutex Links indicate that actions / preconditions **cannot** occur together.  
Mutual Exclusion encodes the constraints on the action space.

## Graph Nodes: $P_i$ – Literals, $A_i$ Actions

Add an action in level  $A_i$  if all of its preconditions are in  $P_i$

Add a precondition in  $P_{i+1}$  if it is the effect of some action in  $A_i$

### Mutex Actions:

- One “clobbers” the other’s effects or preconditions
- They have Mutex preconditions

### Mutex Preconditions:

- All ways of achieving them are mutex

Mutual Exclusion encodes the constraints on the action space.

## Example: Dinner Date

- Init: Garbage CleanHands Quiet
- Goal: Dinner Present  $\neg$  Garbage
- Actions:

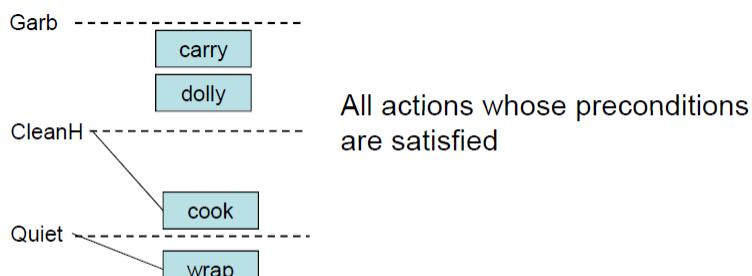
cook	PC: CleanHands	EF: Dinner
wrap	PC: Quiet	EF: Present
carry	PC:	EF: $\neg$ Garbage $\neg$ CleanHands
dolly	PC:	EF: $\neg$ Garbage $\neg$ Quiet

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33

## Initial Actions: $A_0$



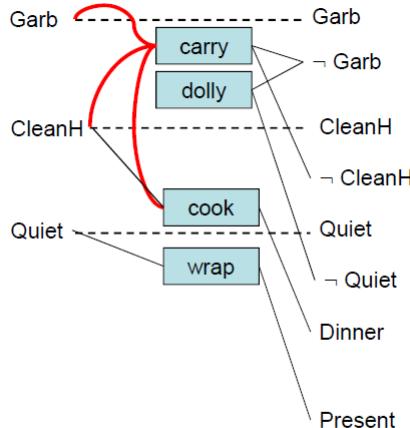
Init:	Garbage	CleanHands	Quiet
Goal:	Dinner	Present	$\neg$ Garbage
Actions:			
cook	PC: CleanHands	EF: Dinner	
wrap	PC: Quiet	EF: Present	
carry	PC:	EF: $\neg$ Garbage $\neg$ CleanHands	
dolly	PC:	EF: $\neg$ Garbage $\neg$ Quiet	

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34

## A<sub>0</sub> Mutex: CARRY Action



Mutex Actions:

- One “clobbers” the other’s effects or preconditions

- They have Mutex preconditions

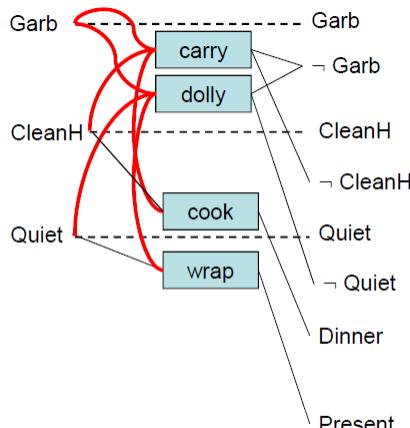
Init:	Garbage	Goal:	CleanHands	Present	Quiet
Actions:					¬ Garbage
cook	PC: CleanHands			EF: Dinner	
wrap	PC: Quiet			EF: Present	
carry	PC:		EF: ¬ Garbage	¬ CleanHands	
dolly	PC:		EF: ¬ Garbage	¬ Quiet	

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35

## A<sub>0</sub> Mutex: All



Mutex Actions:

- One “clobbers” the other’s effects or preconditions

- They have Mutex preconditions

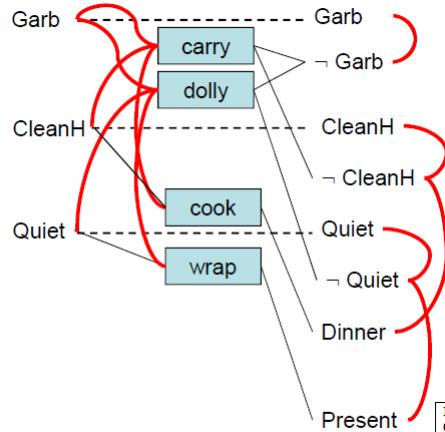
Init:	Garbage	Goal:	CleanHands	Present	Quiet
Actions:					¬ Garbage
cook	PC: CleanHands			EF: Dinner	
wrap	PC: Quiet			EF: Present	
carry	PC:		EF: ¬ Garbage	¬ CleanHands	
dolly	PC:		EF: ¬ Garbage	¬ Quiet	

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36

## A<sub>0</sub> Mutex: Complete



### Mutex Actions:

- One "clobbers" the other's effects or preconditions

- They have Mutex preconditions

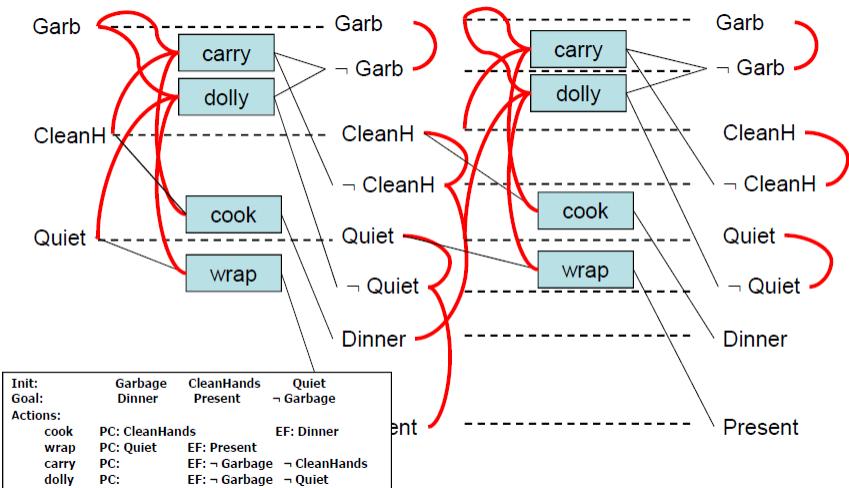
Init:	Garbage	Goal:	CleanHands	Present	Quiet
Actions:					
cook	PC: CleanHands			EF: Dinner	
wrap	PC: Quiet			EF: Present	
carry	PC:		EF: ~Garbage	~CleanHands	
dolly	PC:		EF: ~Garbage	~Quiet	

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37

## A<sub>1</sub>: Action P<sub>2</sub>: Precondition



Init:	Garbage	Goal:	CleanHands	Present	Quiet
Actions:					
cook	PC: CleanHands			EF: Dinner	
wrap	PC: Quiet			EF: Present	
carry	PC:		EF: ~Garbage	~CleanHands	
dolly	PC:		EF: ~Garbage	~Quiet	

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38

## When can we make a plan?

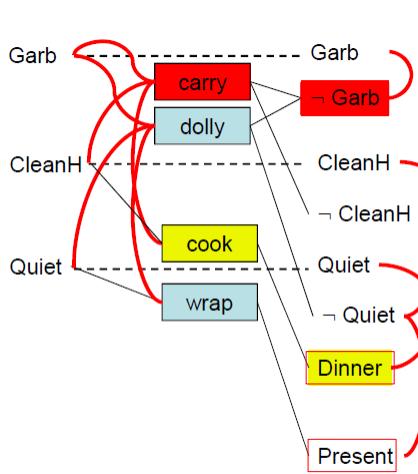
- All goal conditions are satisfied
- Actions at the same level don't interfere
- Each action's preconditions are made true
- Is this sufficient?
  - No, but it's necessary
- Regression search from goals to check for Mutex Conditions

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39

## Solution at Level 1?



- Build a graph
  - Extract a solution(Backward Search)
- Repeat until goal set is empty  
If goals are present & non-mutex:  
1) Choose set of non-mutex actions to achieve each goal  
2) Add preconditions to next goal set

Tow possible sets of actions at level 1:  
{wrap, cook, dolly}  
{wrap, cook, carry}

Neither set works – why?

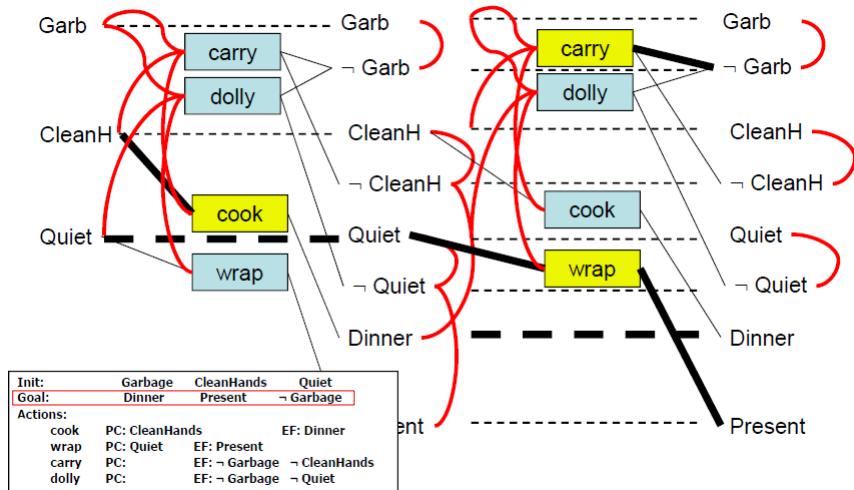
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carry	PC:	EF: $\neg$ Garbage	$\neg$ CleanHands
dolly	PC:	EF: $\neg$ Garbage	$\neg$ Quiet

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40

## Add new layer

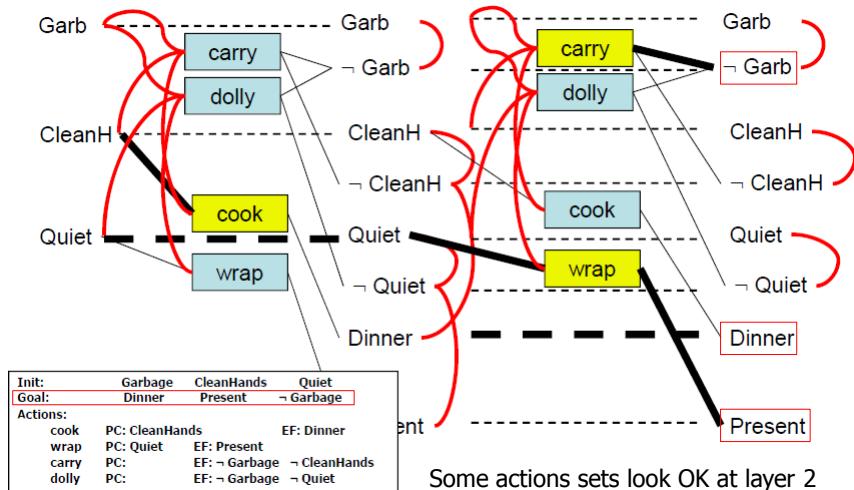


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41

## Solution at Level 2?

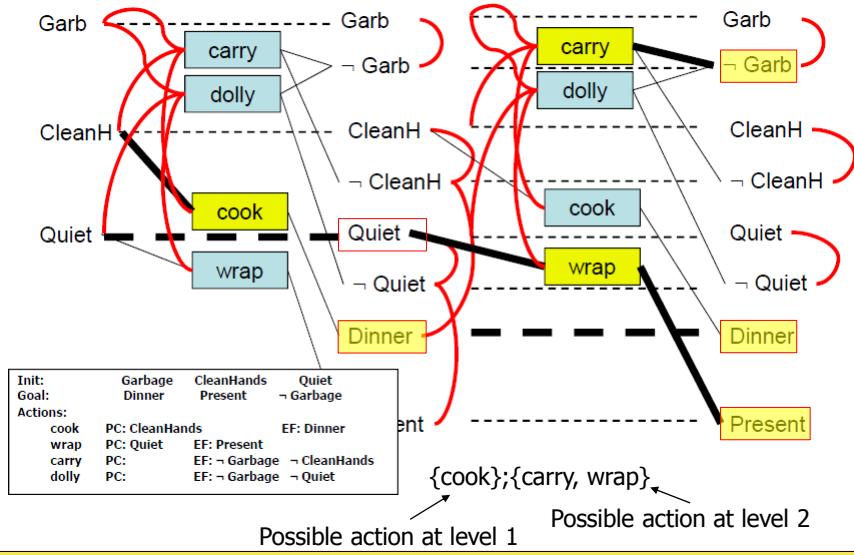


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42

## Solution at Level 2?

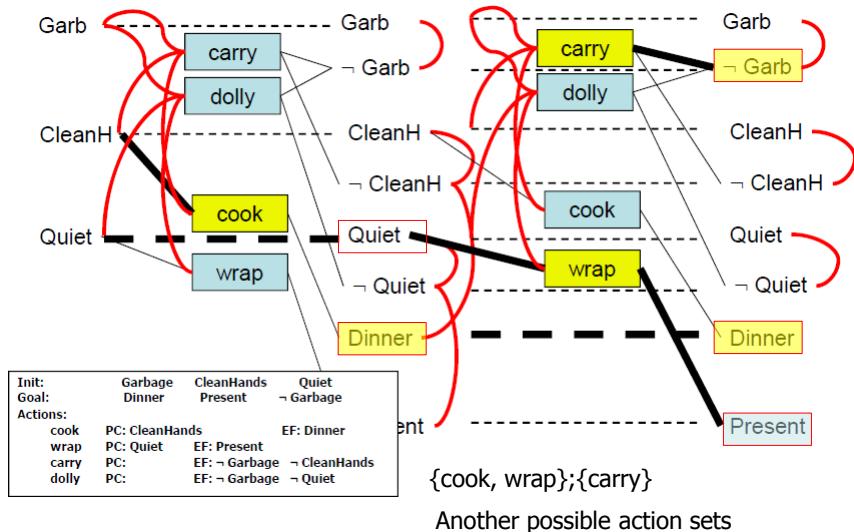


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43

## Solution at Level 2?



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44

## Observations

- Actions monotonically increase
- Propositions monotonically increase
- Mutex Conditions monotonically decrease

## Graphplan Conclusions

- Combines simplicity of state-space planning with constraints of partial order planning
- Very popular method for extending and evaluating improvements

	<b>State Space (60s - )</b>	<b>Plan Space (70s - )</b>	<b>Graph Space (90s - )</b>
<b>Algorithm</b>	Progression Regression	Partial Order Planning	Iterates Graph Building Regression Search
<b>Nodes</b>	World States	Partial Plans	Graph Levels
<b>Transitions</b>	Actions  -Move(x,y,z) -Load(x,y) -Open(r)	Plan Refinement Operations  -Adding Steps -Promotion -Demotion	Sets of Actions  -Constraint Coding -Mutual Exclusion

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9/2/2014

47

<b>Modern Action Representation</b>
<pre> Move ?b from ?x to ?y parameters: ?b, ?x, ?y preconditions: (and (on ?b ?x) (clear ?b) (clear ?y)                   (= ?b ?x) (= ?b ?y) (= ?x ?y)                   (= ?y Table)) effects: (and (on ?b ?y) (not (on ?b ?x)) (clear ?x) (not (clear ?y))) </pre>
PDDL: Planning Domain Definition Language*
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48

## Expressive Actions

- Actions with variables
- Disjunctive preconditions
- Conditional effects
- Universal quantification

### UCPOP [Penberthy and Weld'92]

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9/2/2014

49

## Action Representation

Move ?b from ?x to ?y  
parameters: ?b, ?x, ?y  
preconditions: (and (on ?b ?x) (clear ?b) (clear ?y)  
(≠ ?b ?x) (≠ ?b ?y) (≠ ?x ?y)  
(≠ ?y Table))  
effects: (and (on ?b ?y) (not (on ?b ?x)) (clear ?x) (not (clear ?y)))

vs

Move ?b from ?x to ?y  
parameters: ?b, ?x, ?y  
preconditions: (and (on ?b ?x) (clear ?b) (clear ?y)  
(≠ ?b ?x) (≠ ?b ?y) (≠ ?x ?y)  
effects: (and (on ?b ?y) (not (on ?b ?x)) (clear ?x)  
(when (= ?y Table) (not (clear ?y))))

Conditional Effect



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9/2/2014

50

## Conditional Planning

- Conditional Effects
  - Effects that depend on state
  - Require special attention but same planning concept
- **Disjunctive Effects**
  - Different concept entirely
  - We don't know the outcome of an action

