

# CS 1571 Introduction to AI

## Lecture 21

### Planning: STRIPS

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### Planning

#### Planning problem:

- find a sequence of actions that achieves some goal
- An instance of a search problem

#### Methods for modeling and solving a planning problem:

- State space search
- Situation calculus based on FOL
  - Inference rules
  - Resolution refutation

## Planning problems

### Properties of (real-world) planning problems:

- The description of the **state of the world is very complex**
- **Many possible actions** to apply in any step
- **Actions are typically local**
  - - they affect only a small portion of a state description
- **Goals** are defined as conditions and **refer only to a small portion of state**
- Plans consists of a **long sequence of actions**
- The state space search and situation calculus frameworks may be too cumbersome and inefficient to represent and solve the planning problems

## Situation calculus: problems

**Frame problem** refers to:

- The need to represent a large number of frame axioms

**Solution:** combine positive and negative effects in one rule

$$On(u, v, DO(MOVE(x, y, z), s)) \Leftrightarrow (\neg((u = x) \wedge (v = y)) \wedge On(u, v, s)) \vee \\ \vee (((u = x) \wedge (v = z)) \wedge On(x, y, s) \wedge Clear(x, s) \wedge Clear(z, s))$$

**Inferential frame problem:**

- We still need to derive properties that remain unchanged

**Other problems:**

- **Qualification problem** – enumeration of all possibilities under which an action holds
- **Ramification problem** – enumeration of all inferences that follow from some facts

## Solutions

- **Complex state description and local action effects:**
  - avoid the enumeration and inference of every state component, focus on changes only
- **Many possible actions:**
  - Apply actions that make progress towards the goal
  - Understand what the effect of actions is and reason with the consequences
- **Sequences of actions in the plan can be too long:**
  - Many goals consists of independent or nearly independent sub-goals
  - Allow goal decomposition & divide and conquer strategies

## STRIPS planner

Defines a **restricted representation language** as compared to the situation calculus

**Advantage:** leads to more efficient planning algorithms.

- State-space search with structured representations of states, actions and goals
- Action representation avoids the frame problem

**STRIPS planning problem:**

- much like a standard search (planning) problem;

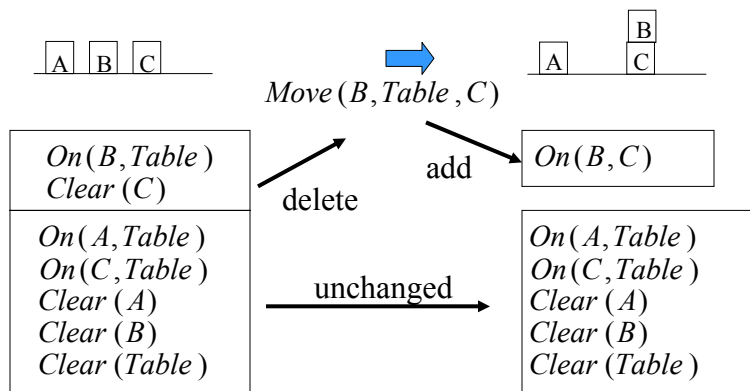
## STRIPS planner

- **States:**
  - conjunction of literals, e.g.  $On(A,B)$ ,  $On(B,Table)$ ,  $Clear(A)$
  - represent facts that are true at a specific point in time
- **Actions (operators):**
  - **Action:**  $Move(x,y,z)$
  - **Preconditions:** conjunctions of literals with variables  
 $On(x,y)$ ,  $Clear(x)$ ,  $Clear(z)$
  - **Effects.** Two lists:
    - **Add list:**  $On(x,z)$ ,  $Clear(y)$
    - **Delete list:**  $On(x,y)$ ,  $Clear(z)$
    - Everything else remains untouched (is preserved)

## STRIPS planning

**Operator:**  $Move(x,y,z)$

- **Preconditions:**  $On(x,y)$ ,  $Clear(x)$ ,  $Clear(z)$
- **Add list:**  $On(x,z)$ ,  $Clear(y)$
- **Delete list:**  $On(x,y)$ ,  $Clear(z)$



## STRIPS planning

### Initial state:

- Conjunction of literals that are true

### Goals in STRIPS:

- A goal is a partially specified state
- Is defined by a conjunction of ground literals
  - No variables allowed in the description of the goal

Example:

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

## Search in STRIPS

### Objective:

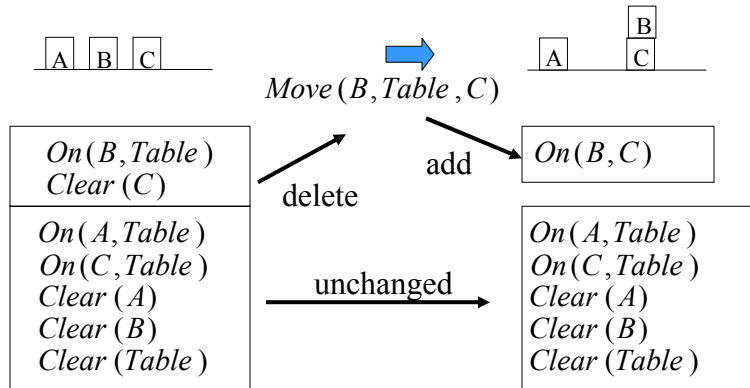
**Find a sequence of operators (a plan) from the initial state to the state satisfying the goal**

**Two approaches** to build a plan:

- **Forward state space search (goal progression)**
  - Start from what is known in the initial state and apply operators in the order they are applied
- **Backward state space search (goal regression)**
  - Start from the description of the goal and identify actions that help to reach the goal

## Forward search (goal progression)

- **Idea:** Given a state  $s$ 
  - Unify the preconditions of some operator  $a$  with  $s$
  - Add and delete sentences from the add and delete list of an operator  $a$  from  $s$  to get a new state



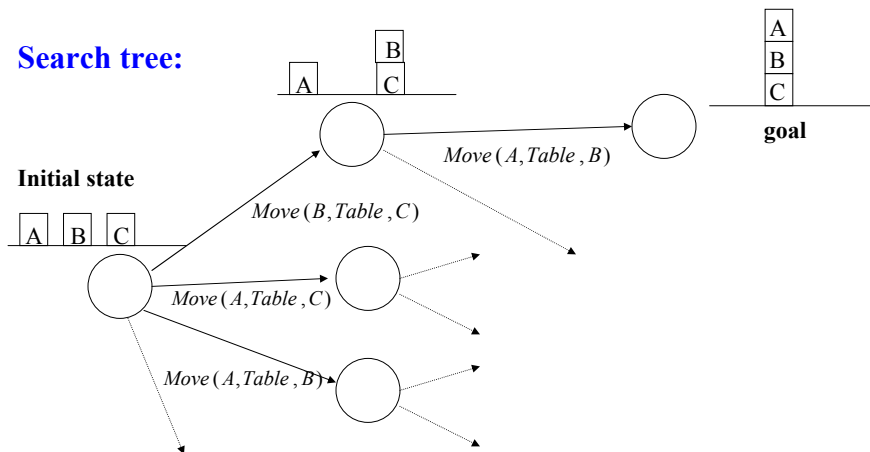
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## Forward search (goal progression)

- Use operators to generate new states to search
- Check new states whether they satisfy the goal

### Search tree:



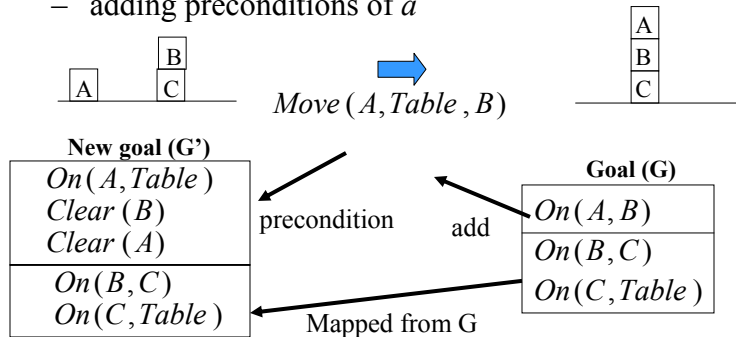
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## Backward search (goal regression)

**Idea:** Given a goal  $G$

- Unify the add list of some operator  $a$  with a subset of  $G$
- If the delete list of  $a$  does not remove elements of  $G$ , then the goal regresses to a new goal  $G'$  that is obtained from  $G$  by:
  - deleting add list of  $a$
  - adding preconditions of  $a$



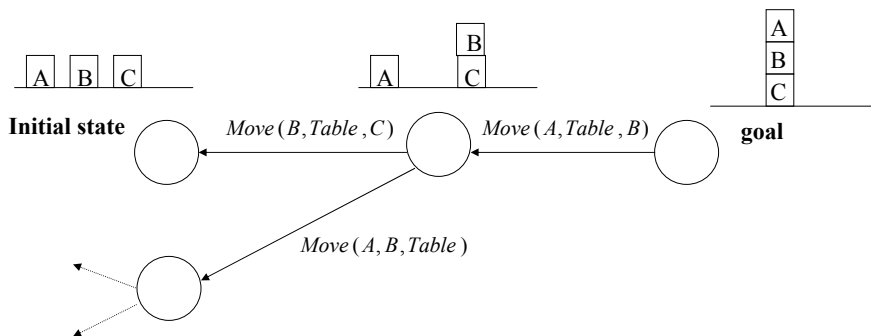
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## Backward search (goal regression)

- Use operators to generate new goals
- Check whether the initial state satisfies the goal

**Search tree:**



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## State-space search

- **Forward and backward state-space planning approaches:**
  - Work with strictly linear sequences of actions
- **Disadvantages:**
  - They cannot take advantage of the **problem decompositions** in which the goal we want to reach consists of a set of independent or nearly independent sub-goals
  - Action sequences cannot be **built from the middle**
  - No mechanism to represent **least commitment** in terms of the action ordering

## Divide and conquer

- **Divide and conquer strategy:**
  - divide the problem to a set of smaller sub-problems,
  - solve each sub-problem independently
  - combine the results to form the solution

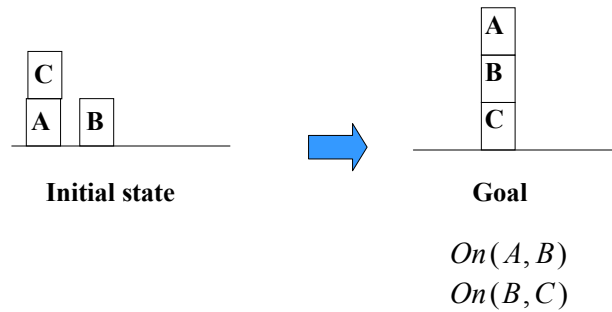
In planning we would like to satisfy a set of goals

- **Divide and conquer in planning:**
  - Divide the planning goals along individual goals
  - Solve (find a plan for) each of them independently
  - Combine the plan solutions in the resulting plan
- Is it always safe to use divide and conquer?
  - No. There can be interacting goals.



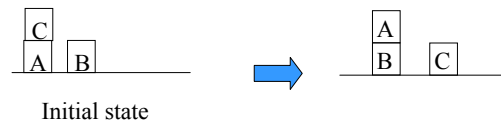
## Sussman's anomaly.

- An example from the blocks world in which the divide and conquer fails due to interacting goals



## Sussman's anomaly

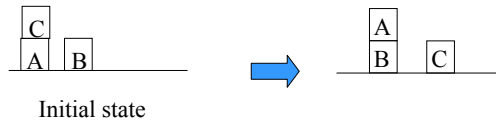
- Assume we want to satisfy  $On(A, B)$  first



But now we cannot satisfy  $On(B, C)$  without undoing  $On(A, B)$

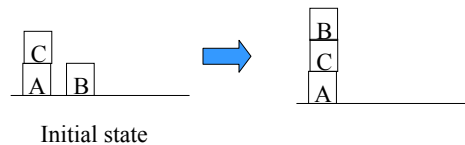
## Sussman's anomaly

1. Assume we want to satisfy  $On(A, B)$  first



But now we cannot satisfy  $On(B, C)$  without undoing  $On(A, B)$

2. Assume we want to satisfy  $On(B, C)$  first.



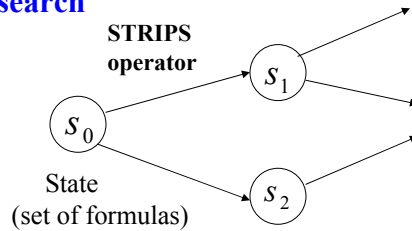
But now we cannot satisfy  $On(A, B)$  without undoing  $On(B, C)$

## State space vs. plan space search

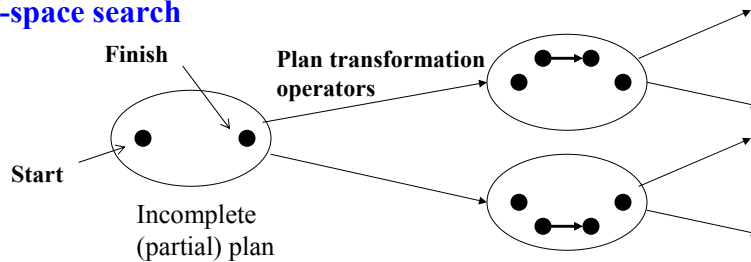
- An alternative to planning algorithms that search states (configurations of world)
- **Plan:** Defines a sequence of operators to be performed
- **Partial plan:**
  - plan that is not complete
    - Some plan steps are missing
  - some orderings of operators are not finalized
    - Only relative order is given
- **Benefits of working with partial plans:**
  - We do not have to build the sequence from the initial state or the goal
  - We do not have to commit to a specific action sequence
  - We can work on sub-goals individually (divide and conquer)

## State-space vs. plan-space search

### State-space search



### Plan-space search



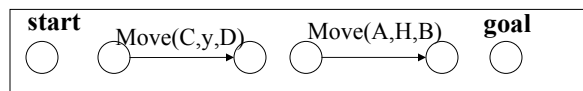
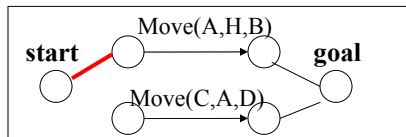
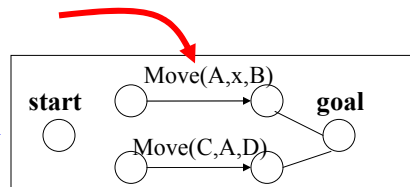
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## Plan transformation operators

Examples of :

- Add an operator to a plan so that it satisfies some open condition
- Add link (+ instantiate)
- Order (reorder) operators



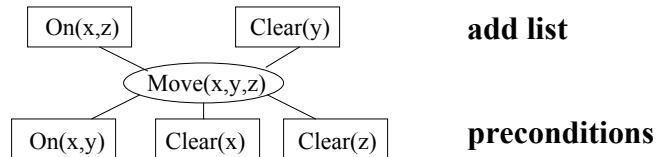
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## Partial-order planners (POP)

- also called **Non-linear planners**
- Use STRIPS operators

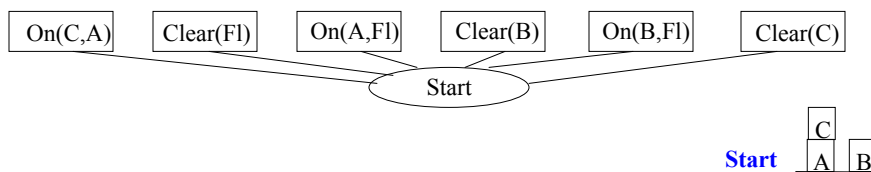
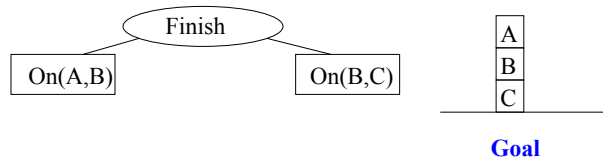
Graphical representation of an **operator** **Move(x,y,z)**



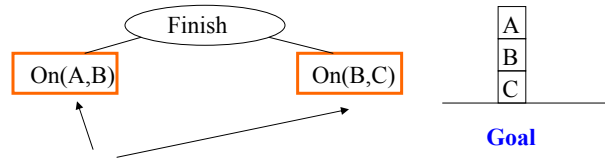
**Delete list is not shown !!!**

Illustration of a POP on the Sussman's anomaly case

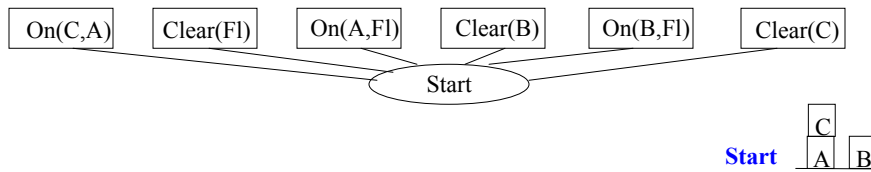
## Partial order planning. Start and finish.



## Partial order planning. Start and finish.



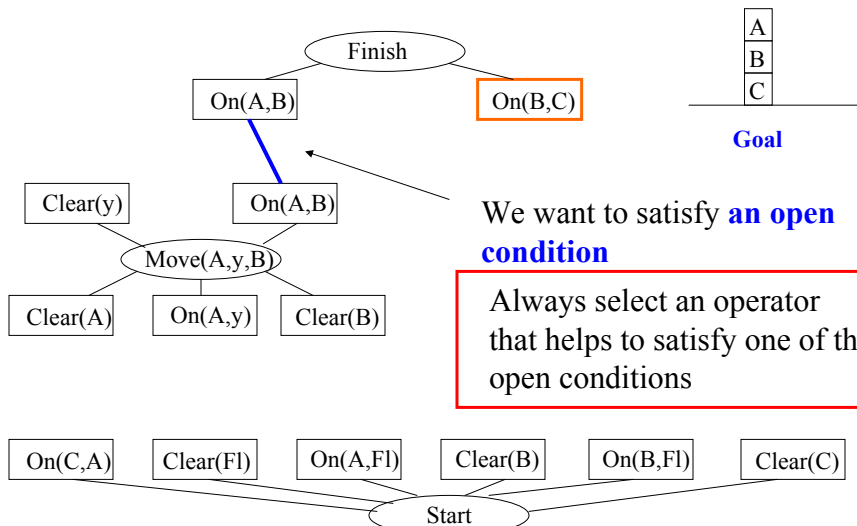
**Open conditions:** conditions yet to be satisfied



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## Partial order planning. Add operator.



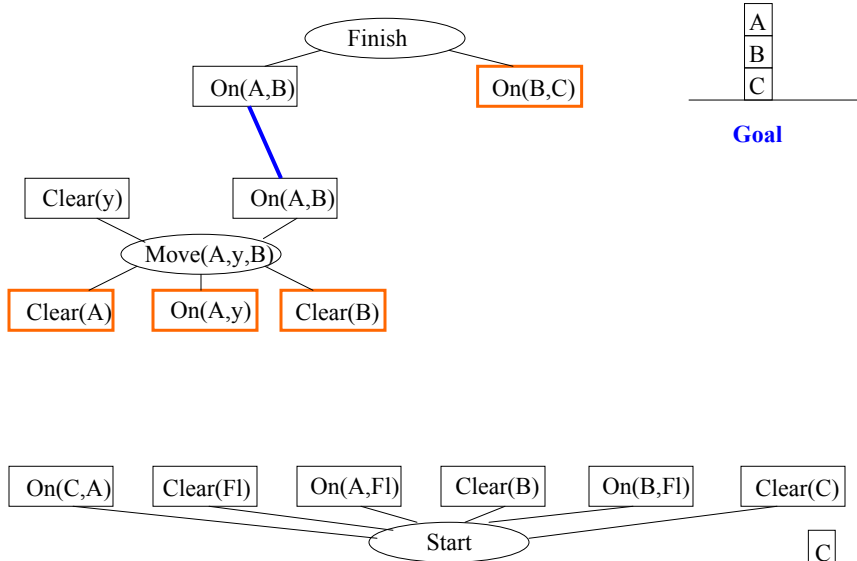
We want to satisfy **an open condition**

Always select an operator that helps to satisfy one of the open conditions

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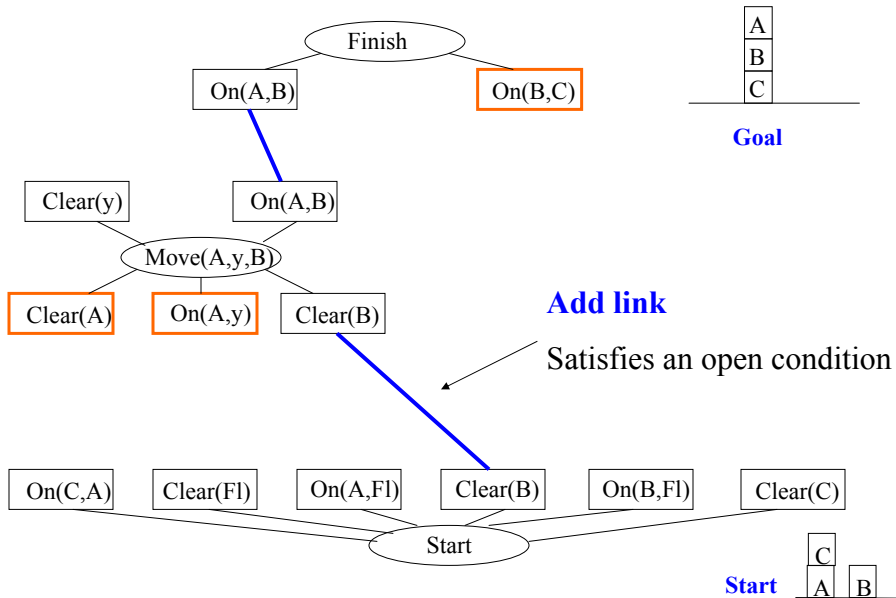
## Partial order planning. Add link.



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## Partial order planning. Add link.



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A
B
C



**Start**

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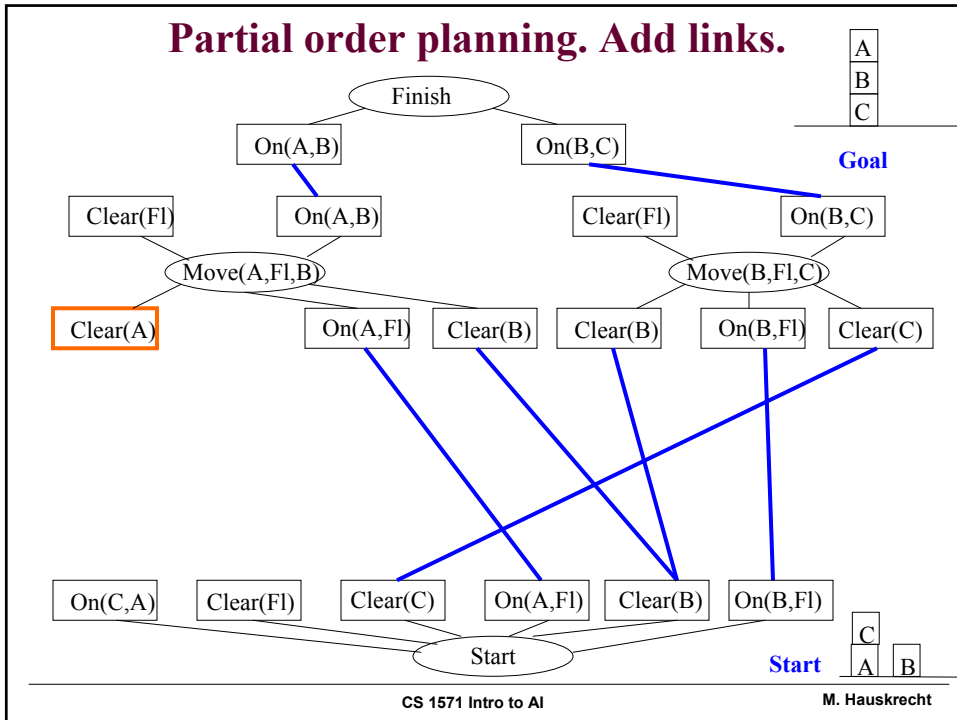
A
B
C



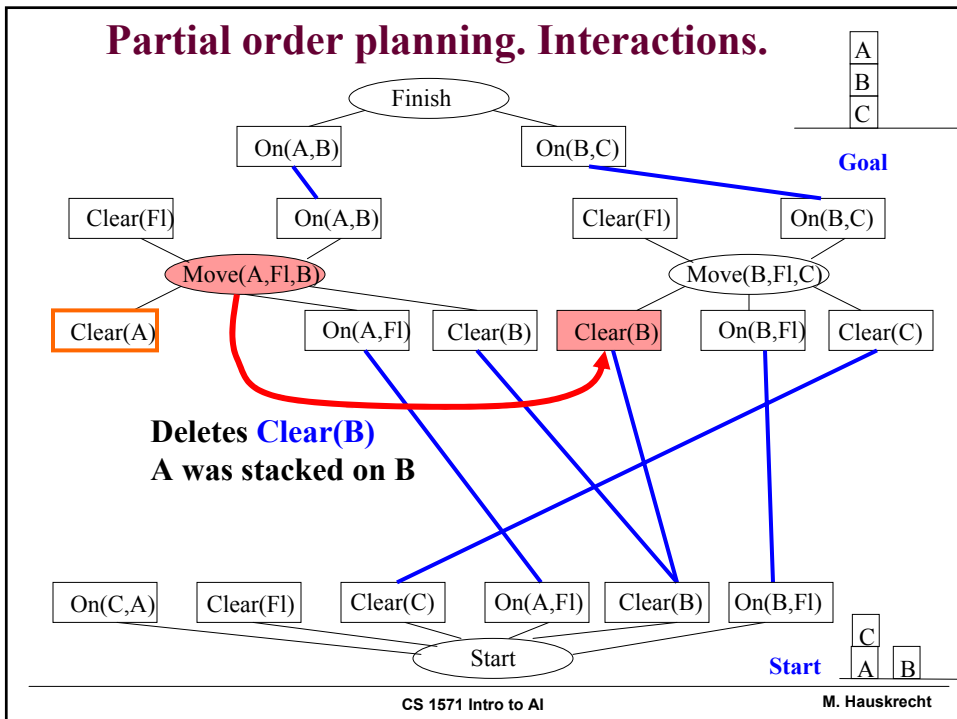
**Start**

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## Partial order planning. Add links.

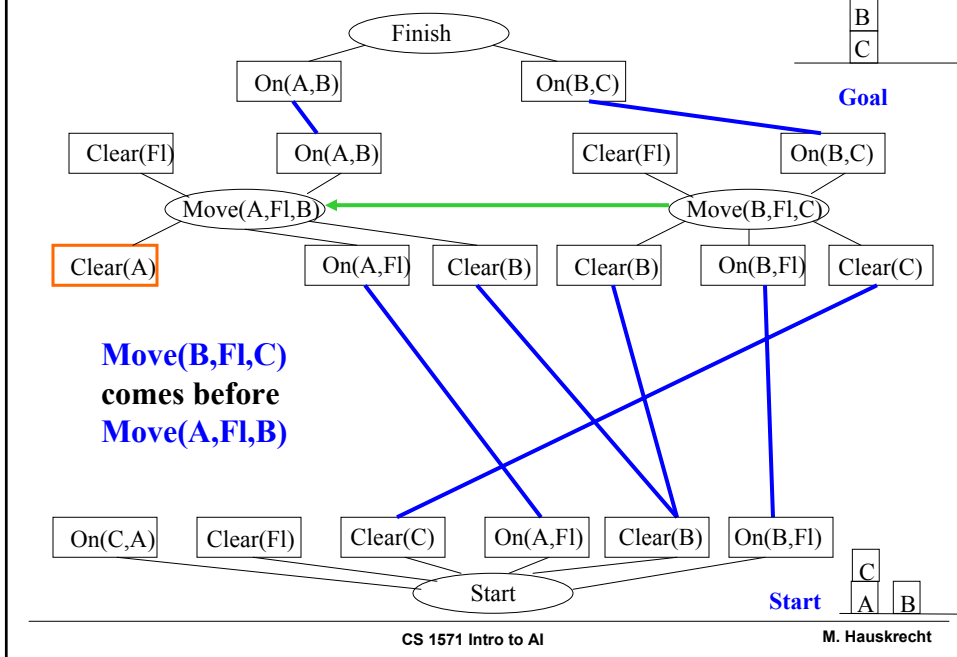


## Partial order planning. Interactions.

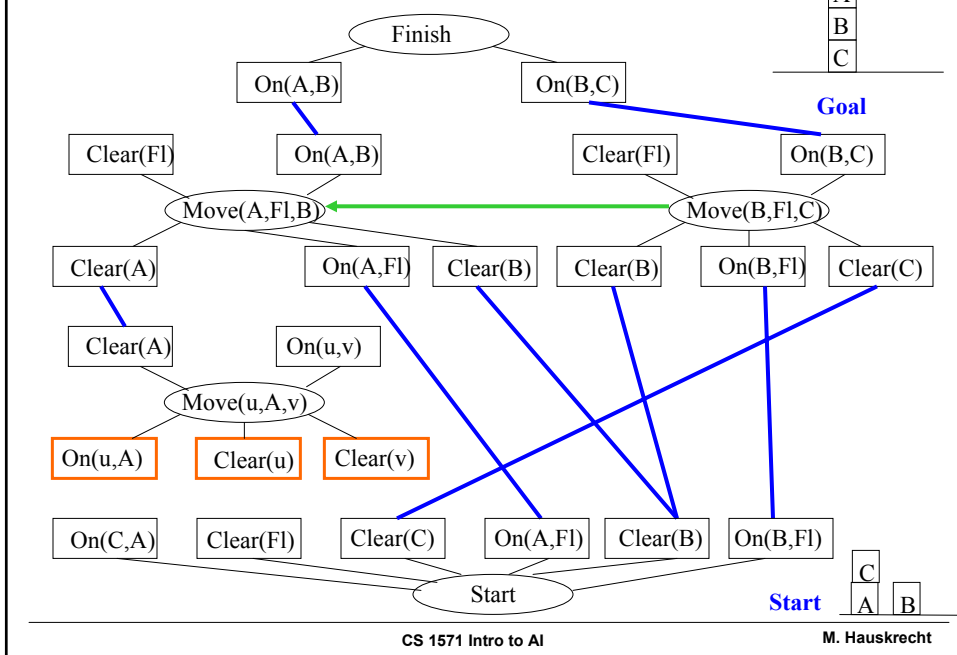




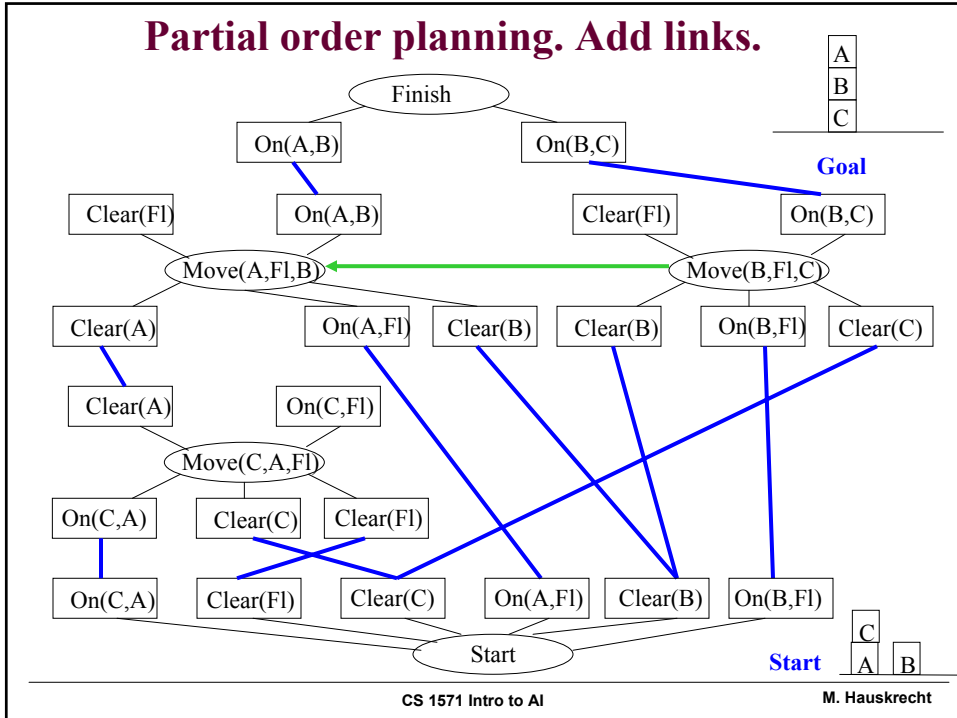
## Partial order planning. Order operators.



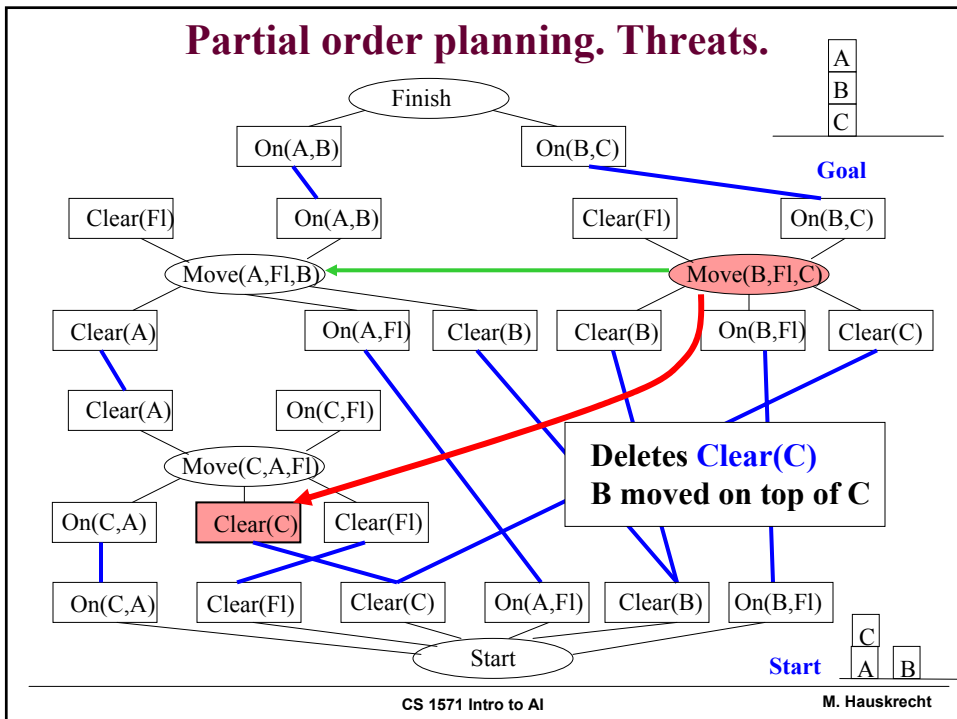
## Partial order planning. Add operator



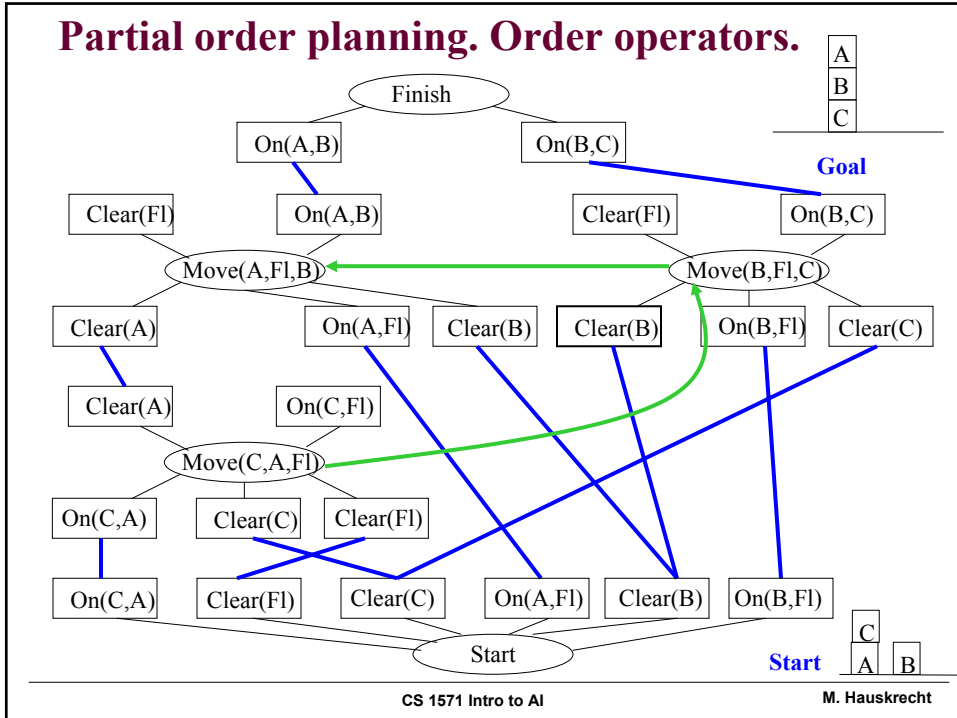
## Partial order planning. Add links.



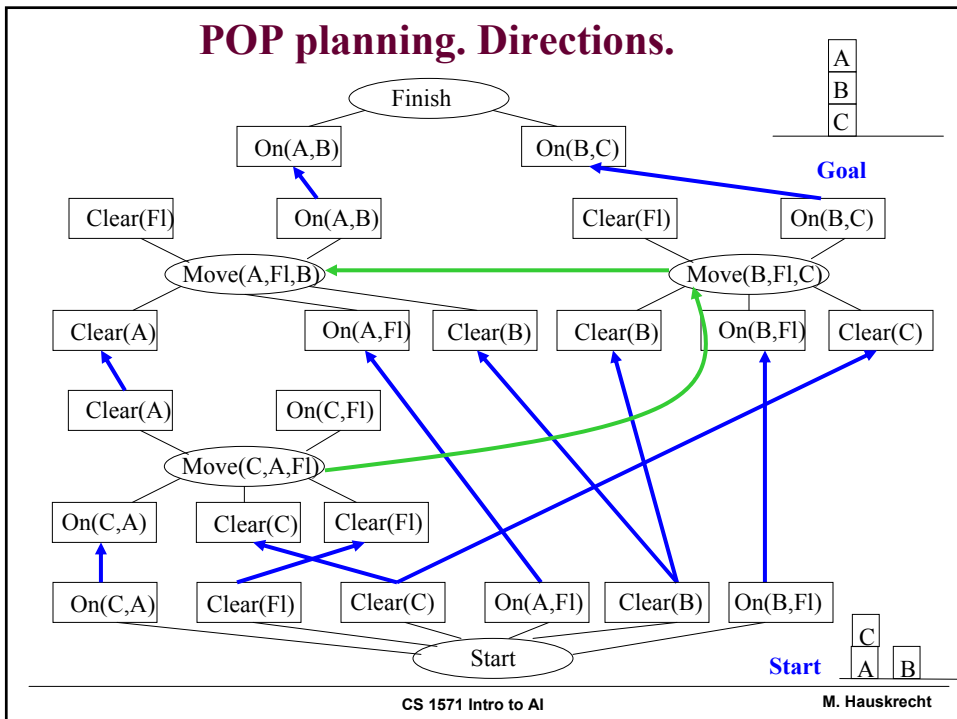
## Partial order planning. Threats.



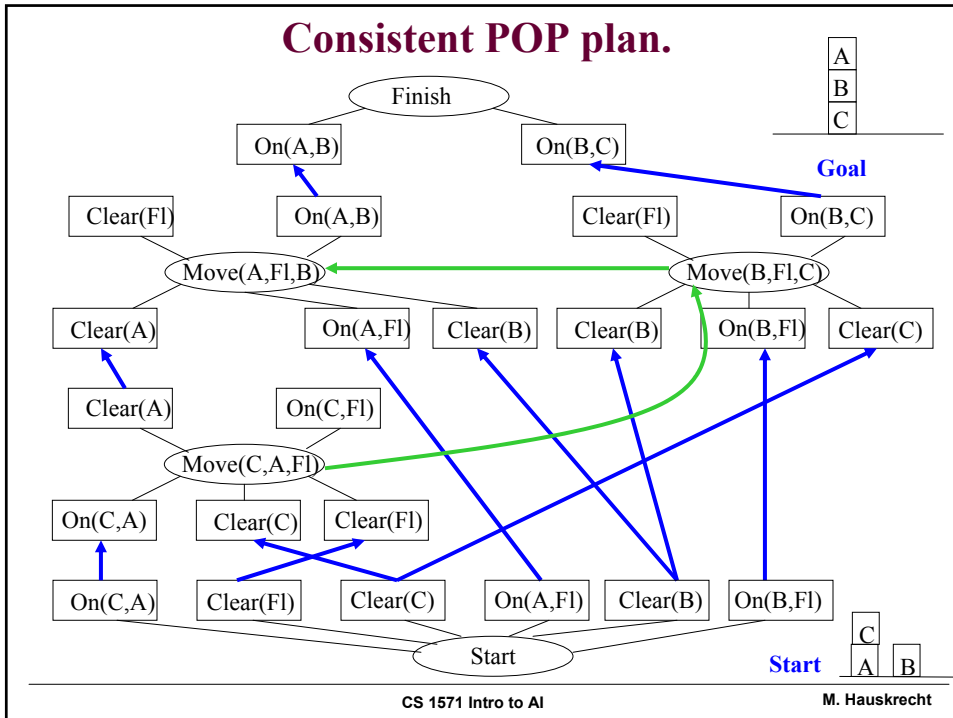
## Partial order planning. Order operators.



## POP planning. Directions.

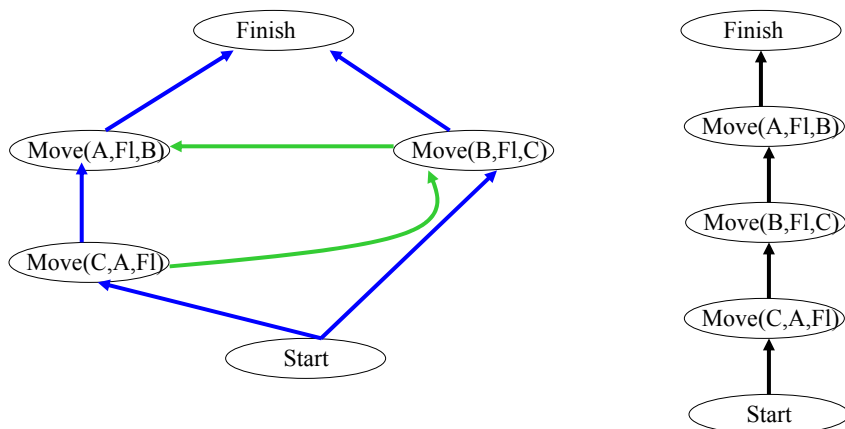


## Consistent POP plan.



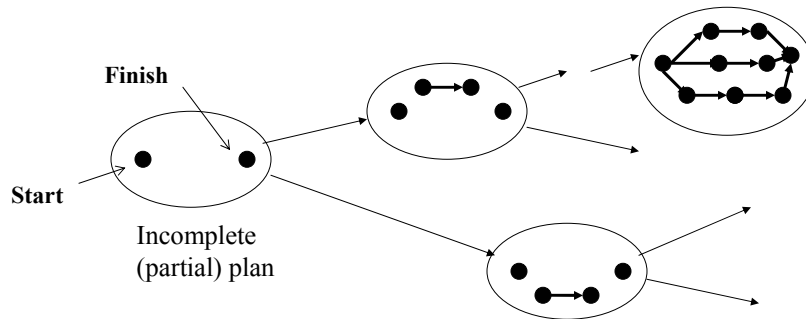
## Partial order planning. Result plan.

Plan: a topological sort of a graph



## Partial order planning.

- **Remember** we search the space of partial plans



- POP: **is sound and complete**