

# Planning

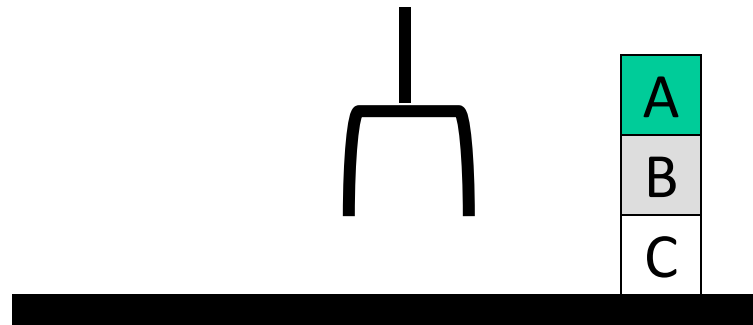
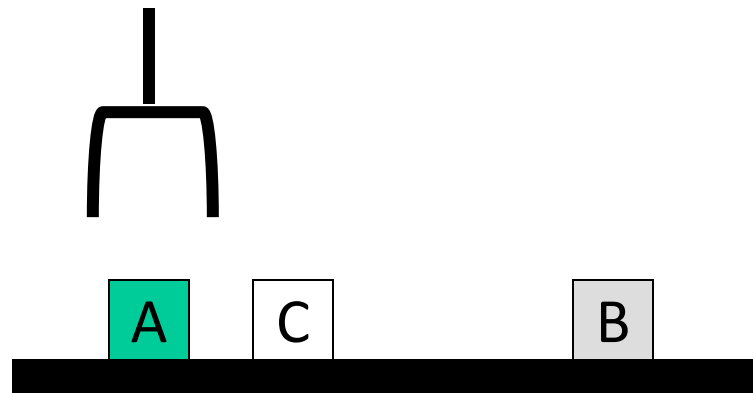
## Chapter 11.1-11.3

Some material adopted from notes  
by Andreas Geyer-Schulz  
and Chuck Dyer

# Overview

- What is planning?
- Approaches to planning
  - GPS / STRIPS
  - Situation calculus formalism [revisited]
  - Partial-order planning

# Blocks World Planning



# Blocks world

The blocks world is a micro-world consisting of a table, a set of blocks and a robot hand

Some domain constraints:

- Only one block can be on another block
- Any number of blocks can be on the table
- The hand can only hold one block

Typical representation uses a logic notation:

ontable(b) ontable(d)

on(c,d) holding(a)

clear(b) clear(c)

# Typical BW planning problem

Initial state:

clear(a)

clear(b)

clear(c)

ontable(a)

ontable(b)

ontable(c)

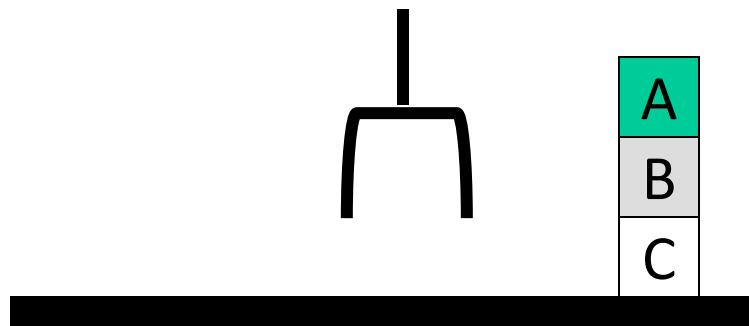
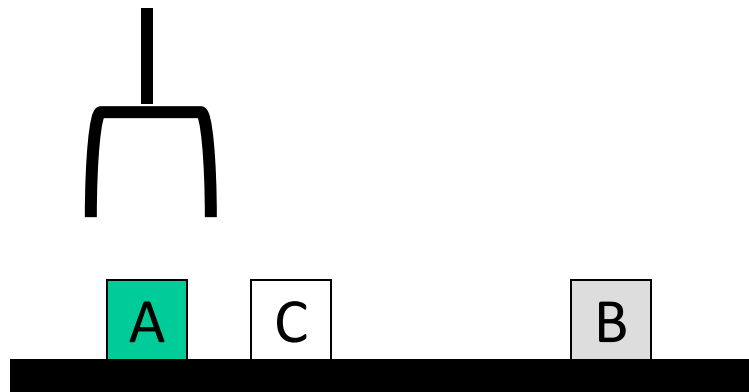
handempty

Goal:

on(b,c)

on(a,b)

ontable(c)



# Typical BW planning problem

Initial state:

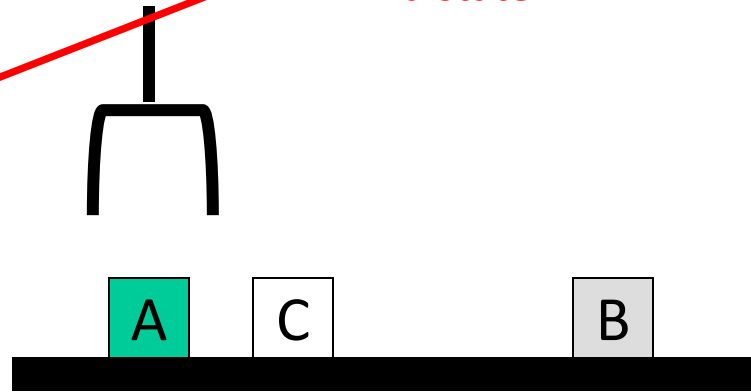
clear(a)  
clear(b)  
clear(c)  
ontable(a)  
ontable(b)  
ontable(c)  
handempty

Goal state:

on(b,c)  
on(a,b)  
ontable(c)

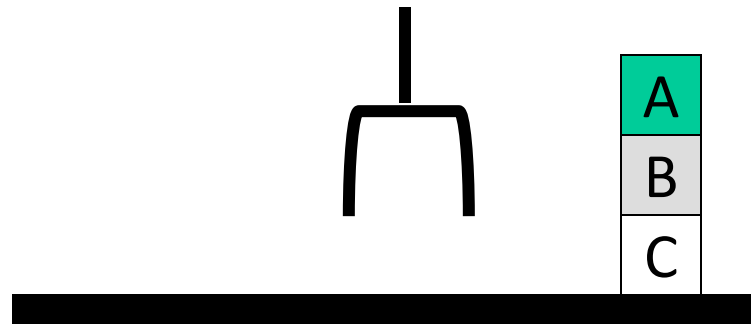
assertions  
describing  
a state

atomic  
robot  
actions



Plan:

pickup(b)  
stack(b,c)  
pickup(a)  
stack(a,b)



# Planning problem

- Find **sequence of actions** to achieve **goal state** when executed from **initial state** given
  - set of possible primitive actions
  - initial state description
  - goal state description or predicate
- compute plan as sequence of action instances that, when executed in initial state, achieves the goal state
- States usually specified as conjunction of conditions, e.g.,  $ontable(a) \wedge on(b, a)$

# Planning vs. problem solving

- Planning and problem solving methods can often solve similar problems
- Planning is more powerful and efficient because of the representations and methods used
- States, goals, and actions are decomposed into sets of sentences (usually in first-order logic)
- Search often proceeds through *plan space* rather than *state space* (though there are also state-space planners)
- Sub-goals can be planned independently, reducing the complexity of the planning problem



# Typical simplifying assumptions

- **Atomic time:** Each action is indivisible
- **No concurrent actions:** but actions need not be ordered w.r.t each other in the plan
- **Deterministic actions:** action results completely determined — no uncertainty in their effects
- Agent is the **sole cause** of change in the world
- Agent is **omniscient** with complete knowledge of the state of the world
- **Closed world assumption:** everything known to be true in world included in state description and anything not listed is false

# Blocks world

The blocks world consists of a table, a set of blocks and a robot hand

Some domain constraints:

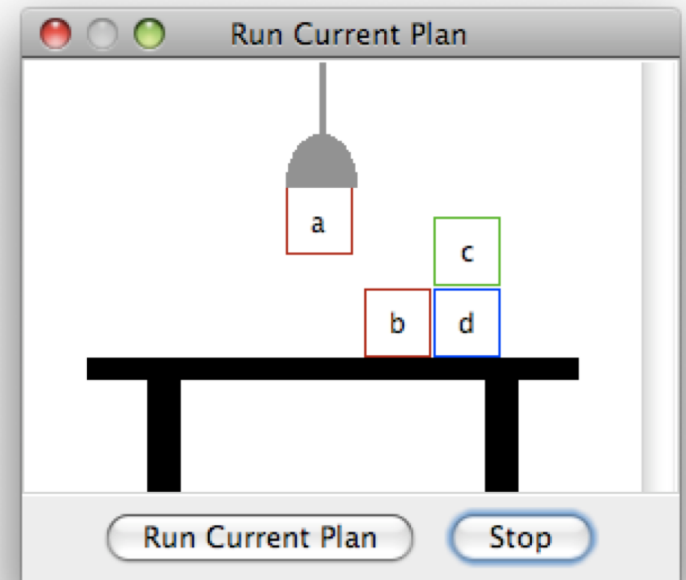
- Only one block can be on another block
- Any number of blocks can be on the table
- The hand can only hold one block

Typical representation:

`ontable(b)` `ontable(d)`

`on(c,d)` `holding(a)`

`clear(b)` `clear(c)`



Meant to be a simple model!

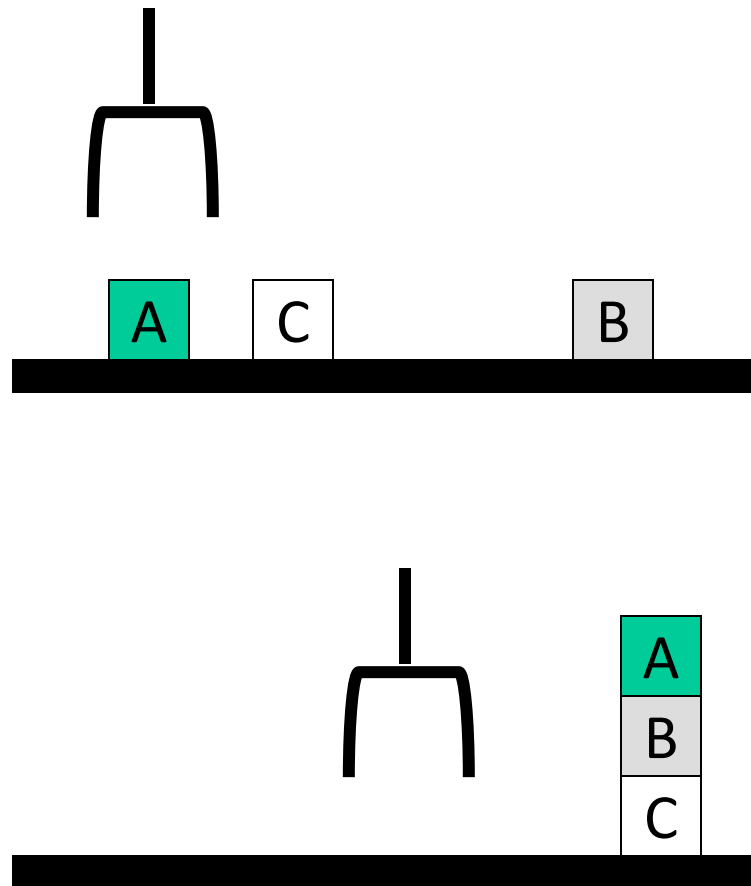
# Typical BW planning problem

Initial state:

clear(a)  
clear(b)  
clear(c)  
ontable(a)  
ontable(b)  
ontable(c)  
handempty

Goal:

on(b,c)  
on(a,b)  
ontable(c)



A plan:

pickup(b)  
stack(b,c)  
pickup(a)  
stack(a,b)

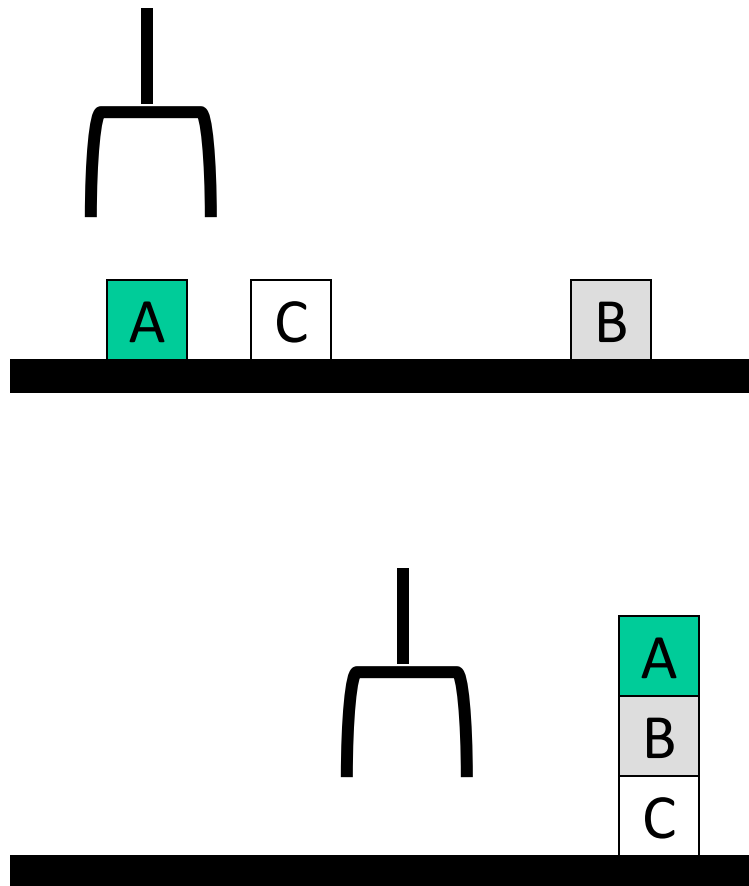
# Another BW planning problem

Initial state:

clear(a)  
clear(b)  
clear(c)  
ontable(a)  
ontable(b)  
ontable(c)  
handempty

Goal:

on(a,b)  
on(b,c)  
ontable(c)



A plan:

pickup(a)  
stack(a,b)  
unstack(a,b)  
putdown(a)  
pickup(b)  
stack(b,c)  
pickup(a)  
stack(a,b)

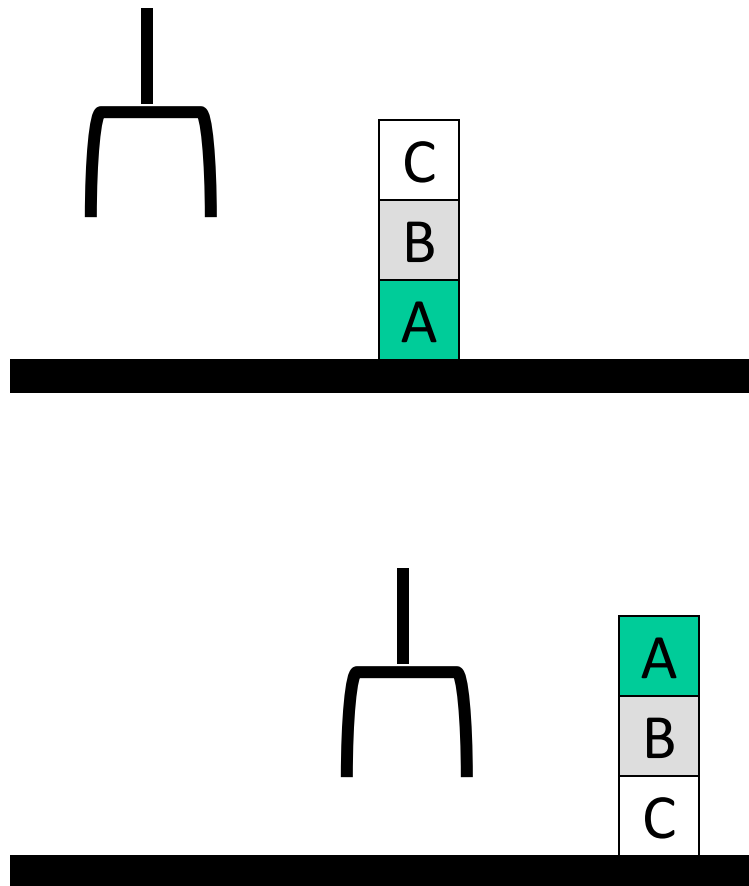
# Yet Another BW planning problem

Initial state:

clear(c)  
ontable(a)  
on(b,a)  
on(c,b)  
handempty

Goal:

on(a,b)  
on(b,c)  
ontable(c)



Plan:

unstack(c,b)  
putdown(c)  
unstack(b,a)  
putdown(b)  
putdown(b)  
pickup(a)  
stack(a,b)  
unstack(a,b)  
putdown(a)  
pickup(b)  
stack(b,c)  
pickup(a)  
stack(a,b)

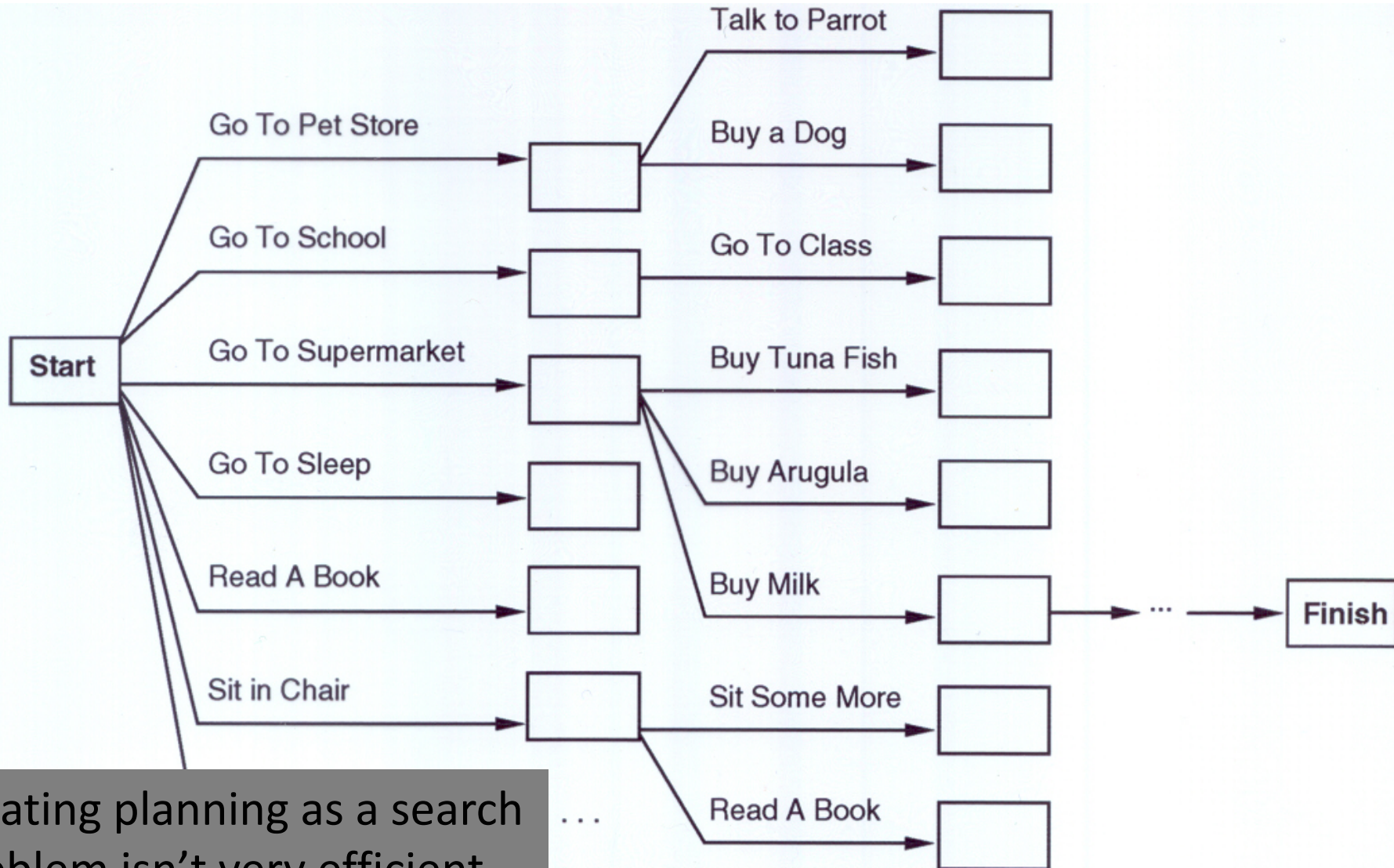
# Major approaches

- Planning as search
- GPS / STRIPS
- Situation calculus
- Partial order planning
- Hierarchical decomposition (HTN planning)
- Planning with constraints (SATplan, Graphplan)
- Reactive planning

# Planning as Search

- Actions: generate successor states
- States: completely described & only used for successor generation, heuristic fn. evaluation & goal testing
- Goals: represented as goal test and using a heuristic function
- Plan representation: unbroken sequences of actions forward from initial states or backward from goal state

“Get a quart of milk, a bunch of bananas and a variable-speed cordless drill.”



Treating planning as a search problem isn't very efficient



# General Problem Solver

- The [General Problem Solver](#) (GPS) system was an early planner (Newell, Shaw, and Simon, 1957)
- GPS generated actions that reduced the difference between some state and a goal state
- GPS used Means-Ends Analysis
  - Compare given to desired states; select best action to do next
  - Table of differences identifies actions to reduce types of differences
- GPS was a state space planner: operated in domain of state space problems specified by initial state, some goal states, and set of operations
- Introduced general way to use domain knowledge to select most promising action to take next



# Situation calculus planning

- Intuition: Represent planning problem using first-order logic
  - Situation calculus lets us reason about changes in the world
  - Use theorem proving to find action sequence, when applied to initial situation leads to desired result
- How “neats” approach the problem

# Situation calculus

- Initial state: logical sentence about (situation)  $S_0$   
 $At(\text{Home}, S_0) \wedge \neg \text{Have}(\text{Milk}, S_0) \wedge \neg \text{Have}(\text{Bananas}, S_0) \wedge \neg \text{Have}(\text{Drill}, S_0)$
- Goal state:  
 $(\exists s) At(\text{Home}, s) \wedge \text{Have}(\text{Milk}, s) \wedge \text{Have}(\text{Bananas}, s) \wedge \text{Have}(\text{Drill}, s)$
- Actions describe how world changes:  
 $\forall (a, s) \text{Have}(\text{Milk}, \text{Result}(a, s)) \Leftrightarrow$   
 $((a = \text{Buy}(\text{Milk}) \wedge At(\text{Grocery}, s)) \vee (\text{Have}(\text{Milk}, s) \wedge a \neq \text{Drop}(\text{Milk})))$
- $\text{Result}(a, s)$  names situation resulting from doing action  $a$  in situation  $s$
- Action sequences:  $\text{Result}'(l, s)$  is result of executing list of actions  $(l)$  starting in  $s$ :  
 $(\forall s) \text{Result}'([], s) = s$   
 $(\forall a, p, s) \text{Result}'([a | p]s) = \text{Result}'(p, \text{Result}(a, s))$

# Situation calculus II

- Solution is plan that when applied to initial state yields situation satisfying the goal:

$At(\text{Home}, \text{Result}'(p, S_0))$

$\wedge \text{Have}(\text{Milk}, \text{Result}'(p, S_0))$

$\wedge \text{Have}(\text{Bananas}, \text{Result}'(p, S_0))$

$\wedge \text{Have}(\text{Drill}, \text{Result}'(p, S_0))$

- We expect a plan (i.e., variable assignment through unification) such as:

$p = [\text{Go}(\text{Grocery}), \text{Buy}(\text{Milk}), \text{Buy}(\text{Bananas}),$   
 $\text{Go}(\text{HardwareStore}), \text{Buy}(\text{Drill}), \text{Go}(\text{Home})]$

# Situation calculus: Blocks world

- A situation calculus rule for the blocks world

Clear (X, Result(A,S))  $\leftrightarrow$

[Clear (X, S)  $\wedge$

( $\neg$ (A=Stack(Y,X)  $\vee$  A=Pickup(X))

$\vee$  (A=Stack(Y,X)  $\wedge$   $\neg$ (holding(Y,S))

$\vee$  (A=Pickup(X)  $\wedge$   $\neg$ (handempty(S)  $\wedge$  ontable(X,S)  $\wedge$  clear(X,S))))]

$\vee$  [A=Stack(X,Y)  $\wedge$  holding(X,S)  $\wedge$  clear(Y,S)]

$\vee$  [A=Unstack(Y,X)  $\wedge$  on(Y,X,S)  $\wedge$  clear(Y,S)  $\wedge$  handempty(S)]

$\vee$  [A=Putdown(X)  $\wedge$  holding(X,S)]

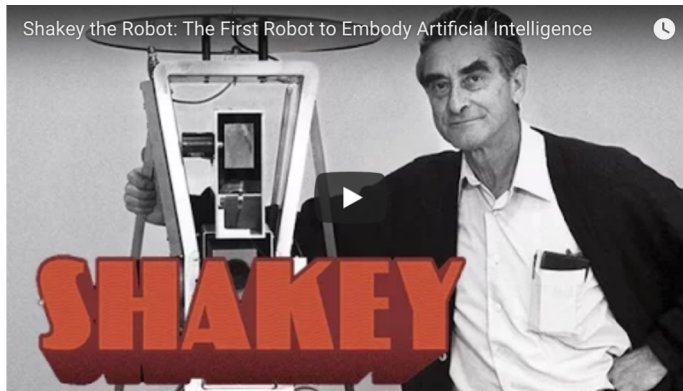
- Translation: A block is clear if (a) in previous state it was clear & we didn't pick it up or stack something on it, or (b) we stacked it on something else, or (c) something was on it that we unstacked, or (d) we were holding it and we put it down.
- Whew!!! There's gotta be a better way!

# Situation calculus planning: Analysis

- Fine in theory, but problem solving (search) is exponential in worst case
- Resolution theorem proving only finds a proof (plan), not necessarily a **good** plan
- So, restrict language and use special-purpose algorithm (a planner) rather than general theorem prover
- Planning is a common task for intelligent agents, so it's reasonable to have a special subsystem for it

# Shakey the robot

First general-purpose mobile robot to be able to reason about its own actions



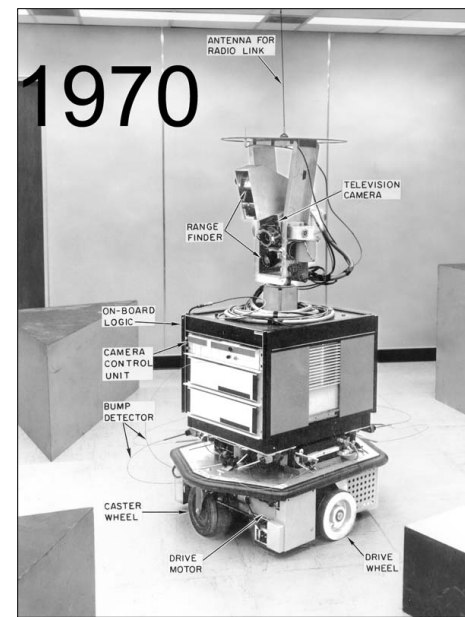
Shakey the Robot: 1st Robot to Embody Artificial Intelligence (2017, 6 min.)



Shakey: Experiments in Robot Planning and Learning (1972, 24 min)

# Strips planning representation

- Classic approach first used in the [STRIPS](#) (Stanford Research Institute Problem Solver) planner
- A State is a conjunction of ground literals  
 $\text{at(Home)} \wedge \neg \text{have(Milk)} \wedge \neg \text{have(bananas)} \dots$
- Goals are conjunctions of literals, but may have variables, assumed to be existentially quantified  
 $\text{at}(?x) \wedge \text{have(Milk)} \wedge \text{have(bananas)} \dots$
- Need not fully specify state
  - Non-specified conditions either don't-care or assumed false
  - Represent many cases in small storage
  - May only represent changes in state rather than entire situation
- Unlike theorem prover, not seeking whether goal is true, but is there a sequence of actions to attain it



[Shakey the robot](#)



# Operator/action representation

- Action operators have three components:

- Action description
- Precondition: conjunction of positive literals
- Effect: conjunction of positive or negative literals describing how situation changes when operator is applied

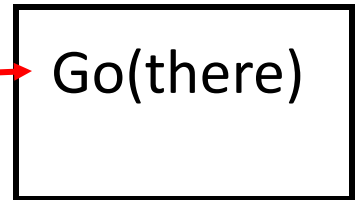
- Example:

Op[Action: Go(there),

Precond:  $\text{At}(\text{here}) \wedge \text{Path}(\text{here}, \text{there})$ ,

Effect:  $\text{At}(\text{there}) \wedge \neg \text{At}(\text{here})$ ]

At(there) , Path(there,here)



At(there) ,  $\neg \text{At}(\text{here})$

- All variables are universally quantified

- Situation variables are implicit

- preconditions must be true in the state immediately before operator is applied; effects are true immediately after

# Blocks world operators

- Classic basic operations for the blocks world
  - `stack(X,Y)`: put block X on block Y
  - `unstack(X,Y)`: remove block X from block Y
  - `pickup(X)`: pickup block X
  - `putdown(X)`: put block X on the table
- Each represented by
  - list of preconditions
  - list of new facts to be added (add-effects)
  - list of facts to be removed (delete-effects)
  - optionally, set of (simple) variable constraints
- For example `stack(X,Y)`:
  - `preconditions(stack(X,Y), [holding(X), clear(Y)])`
  - `deletes(stack(X,Y), [holding(X), clear(Y)])`.
  - `adds(stack(X,Y), [handempty, on(X,Y), clear(X)])`
  - `constraints(stack(X,Y), [X≠Y, Y≠table, X≠table])`

# Blocks world operators (Prolog)

*operator(op, preconditions, adds, deletes, constraints)*

operator(stack(X,Y),  
[holding(X), clear(Y)],  
[handempty, on(X,Y), clear(X)],  
[holding(X), clear(Y)],  
[X≠Y, Y≠table, X≠table]).

operator(unstack(X,Y),  
[on(X,Y), clear(X), handempty],  
[holding(X), clear(Y)],  
[handempty, clear(X), on(X,Y)],  
[X≠Y, Y≠table, X≠table]).

operator(pickup(X),  
[ontable(X), clear(X), handempty],  
[holding(X)],  
[ontable(X), clear(X), handempty],  
[X≠table]).

operator(putdown(X),  
[holding(X)],  
[ontable(X), handempty, clear(X)],  
[holding(X)],  
[X≠table]).

# STRIPS planning

- STRIPS maintains two additional data structures:
  - State List - all currently true predicates.
  - Goal Stack - push down stack of goals to be solved, with current goal on top
- If current goal not satisfied by present state, find operator that adds it and push operator and its preconditions (subgoals) on stack
- When a current goal is satisfied, POP from stack
- When an operator is on top stack, record application of that operator on plan sequence and use operator's add and delete lists to update current state

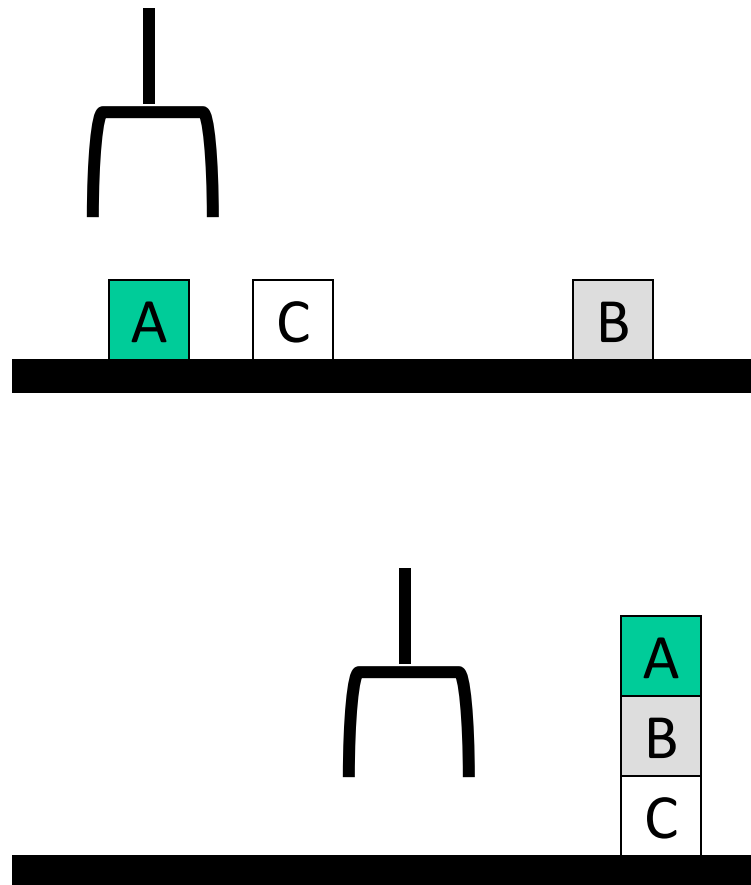
# Typical BW planning problem

Initial state:

clear(a)  
clear(b)  
clear(c)  
ontable(a)  
ontable(b)  
ontable(c)  
handempty

Goal:

on(b,c)  
on(a,b)  
ontable(c)



A plan:

pickup(b)  
stack(b,c)  
pickup(a)  
stack(a,b)

# Strips in Prolog

```
% strips(+Goals, +InitState, -Plan)
strips(Goal, InitState, Plan):-
    strips(Goal, InitState, [], _, RevPlan),
    reverse(RevPlan, Plan).

% strips(+Goals,+State,+Plan,-NewState, NewPlan )
% Finished if each goal in Goals is true
% in current State.
strips(Goals, State, Plan, State, Plan) :-
    subset(Goals,State).
```

```
strips(Goals, State, Plan, NewState, NewPlan):-
    % Goal is an unsatisfied goal.
    member(Goal, Goals),
    (\+ member(Goal, State)),
    % Op is an Operator with Goal as a result.
    operator(Op, Preconditions, Adds, Deletes,_),
    member(Goal,Adds),
    % Achieve the preconditions
    strips(Preconditions, State, Plan, TmpState1,
        TmpPlan1),
    % Apply the Operator
    diff(TmpState1, Deletes, TmpState2),
    union(Adds, TmpState2, TmpState3).
    % Continue planning.
    strips(GoalList, TmpState3, [Op|TmpPlan1],
        NewState, NewPlan).
```

# Trace (Prolog)

strips([on(b,c),on(a,b),ontable(c)],[clear(a),clear(b),clear(c),ontable(a),ontable(b),ontable(c),handempty],[[]])

Achieve on(b,c) via stack(b,c) with preconds: [holding(b),clear(c)]

strips([holding(b),clear(c)],[clear(a),clear(b),clear(c),ontable(a),ontable(b),ontable(c),handempty],[[]])

Achieve holding(b) via pickup(b) with preconds: [ontable(b),clear(b),handempty]

strips([ontable(b),clear(b),handempty],[clear(a),clear(b),clear(c),ontable(a),ontable(b),ontable(c),handempty],[[]])

Applying pickup(b)

strips([holding(b),clear(c)],[clear(a),clear(c),holding(b),ontable(a),ontable(c)],[pickup(b)])

Applying stack(b,c)

strips([on(b,c),on(a,b),ontable(c)],[handempty,clear(a),clear(b),ontable(a),ontable(c),on(b,c)],[stack(b,c),pickup(b)])

Achieve on(a,b) via stack(a,b) with preconds: [holding(a),clear(b)]

strips([holding(a),clear(b)],[handempty,clear(a),clear(b),ontable(a),ontable(c),on(b,c)],[stack(b,c),pickup(b)])

Achieve holding(a) via pickup(a) with preconds: [ontable(a),clear(a),handempty]

strips([ontable(a),clear(a),handempty],[handempty,clear(a),clear(b),ontable(a),ontable(c),on(b,c)],[stack(b,c),pickup(a)])

Applying pickup(a)

strips([holding(a),clear(b)],[clear(b),holding(a),ontable(c),on(b,c)],[pickup(a),stack(b,c),pickup(b)])

Applying stack(a,b)

strips([on(b,c),on(a,b),ontable(c)],[handempty,clear(a),ontable(c),on(a,b),on(b,c)],[stack(a,b),pickup(a),stack(b,c),pickup(b)])

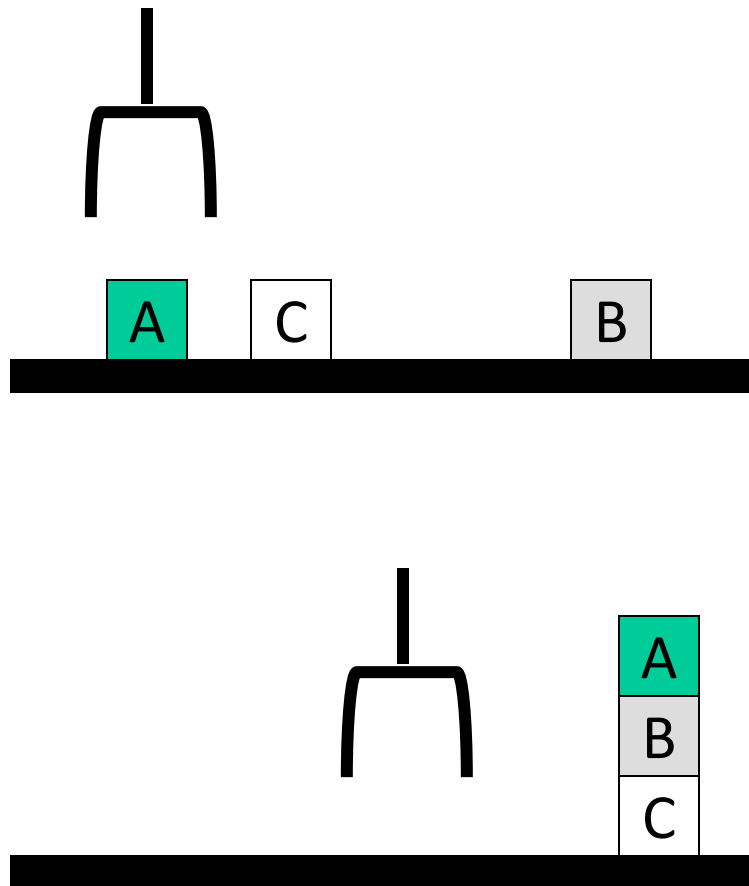
# Another BW planning problem

Initial state:

clear(a)  
clear(b)  
clear(c)  
ontable(a)  
ontable(b)  
ontable(c)  
handempty

Goal:

on(a,b)  
on(b,c)  
ontable(c)



A plan:

pickup(a)  
stack(a,b)  
unstack(a,b)  
putdown(a)  
pickup(b)  
stack(b,c)  
pickup(a)  
stack(a,b)



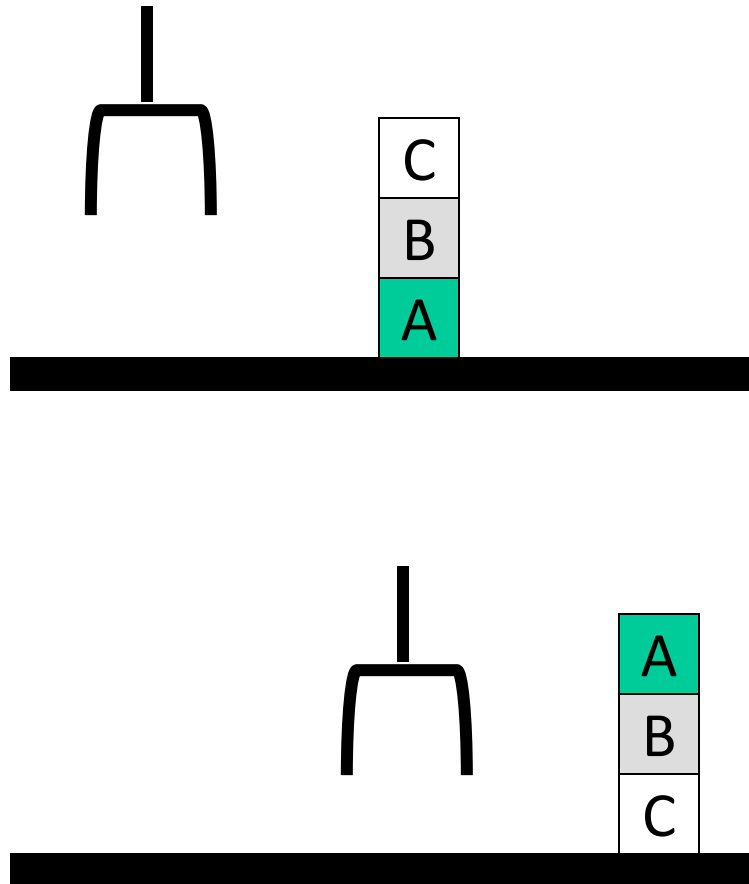
# Yet Another BW planning problem

## Initial state:

clear(c)  
ontable(a)  
on(b,a)  
on(c,b)  
handempty

## Goal:

on(a,b)  
on(b,c)  
ontable(c)



## Plan:

```
unstack(c,b)
putdown(c)
unstack(b,a)
putdown(b)
pickup(b)
stack(b,a)
unstack(b,a)
putdown(b)
pickup(a)
stack(a,b)
unstack(a,b)
putdown(a)
pickup(b)
stack(b,c)
pickup(a)
stack(a,b)
```

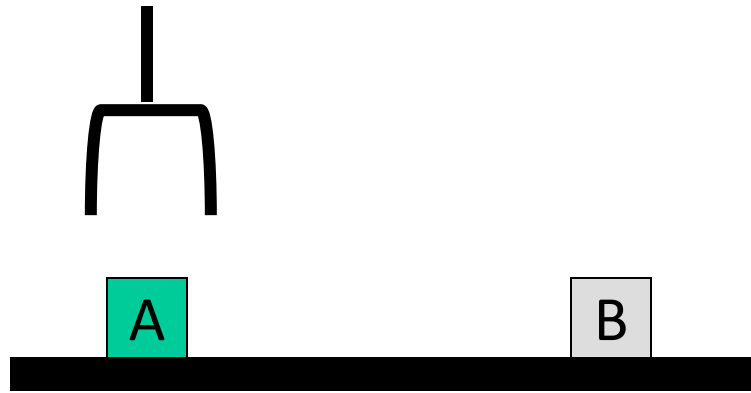
# Yet Another BW planning problem

Initial state:

ontable(a)  
ontable(b)  
clear(a)  
clear(b)  
handempty

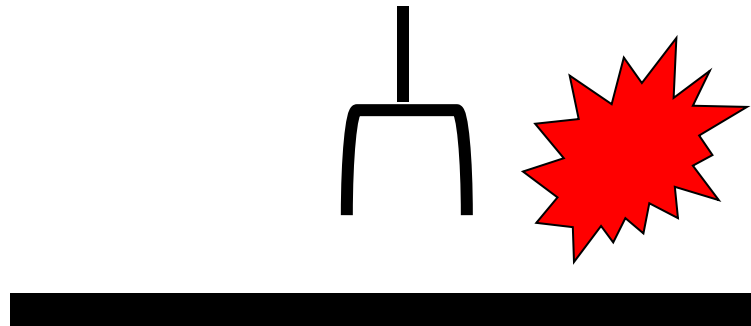
Goal:

on(a,b)  
on(b,a)



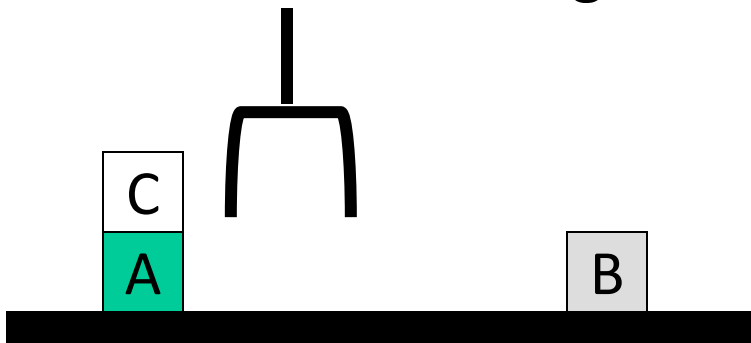
Plan:

??

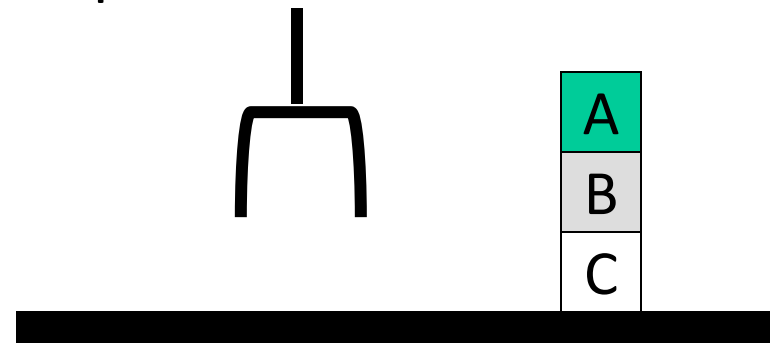


# Goal interaction

- Simple planning algorithms assume independent sub-goals
  - Solve each separately and concatenate the solutions
- The “[Sussman Anomaly](#)” is the classic example of the goal interaction problem:
  - Solving on(A,B) first (via unstack(C,A), stack(A,B)) is undone when solving 2nd goal on(B,C) (via unstack(A,B), stack(B,C))
  - Solving on(B,C) first will be undone when solving on(A,B)
- Classic STRIPS couldn't handle this, although minor modifications can get it to do simple cases



Initial state

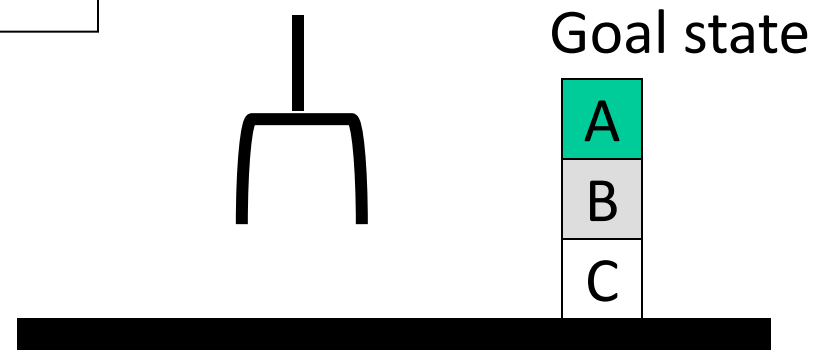
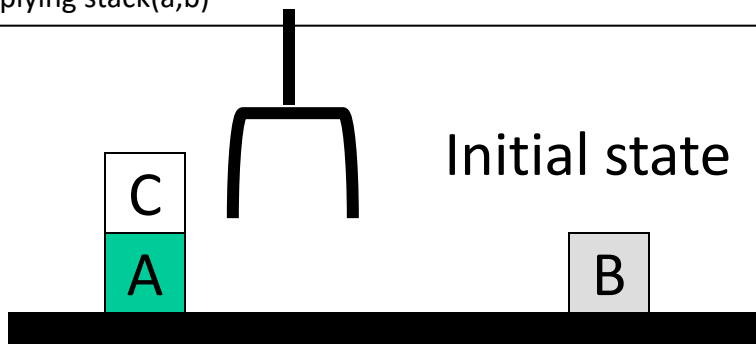


Goal state

# Sussman Anomaly

Achieve on(a,b) via stack(a,b) with preconds: [holding(a),clear(b)]  
| Achieve holding(a) via pickup(a) with preconds: [ontable(a),clear(a),handempty]  
| | Achieve clear(a) via unstack(\_1584,a) with preconds:  
[on(\_1584,a),clear(\_1584),handempty]  
| | Applying unstack(c,a)  
| | Achieve handempty via putdown(\_2691) with preconds: [holding(\_2691)]  
| | Applying putdown(c)  
| Applying pickup(a)  
Applying stack(a,b)  
Achieve on(b,c) via stack(b,c) with preconds: [holding(b),clear(c)]  
| Achieve holding(b) via pickup(b) with preconds: [ontable(b),clear(b),handempty]  
| | Achieve clear(b) via unstack(\_5625,b) with preconds:  
[on(\_5625,b),clear(\_5625),handempty]  
| | Applying unstack(a,b)  
| | Achieve handempty via putdown(\_6648) with preconds: [holding(\_6648)]  
| | Applying putdown(a)  
| Applying pickup(b)  
Applying stack(b,c)  
Achieve on(a,b) via stack(a,b) with preconds: [holding(a),clear(b)]  
| Achieve holding(a) via pickup(a) with preconds: [ontable(a),clear(a),handempty]  
| Applying pickup(a)  
Applying stack(a,b)

From  
[clear(b),clear(c),ontable(a),ontable(b),on(c,a),handempty]  
To [on(a,b),on(b,c),ontable(c)]  
Do:  
unstack(c,a)  
putdown(c)  
pickup(a)  
stack(a,b)  
unstack(a,b)  
putdown(a)  
pickup(b)  
stack(b,c)  
pickup(a)  
stack(a,b)



# Sussman Anomaly

- Classic Strips assumed that once a goal had been satisfied it would stay satisfied
- Simple Prolog version selects any currently unsatisfied goal to tackle at each iteration
- This can handle this problem, at the expense of looping for other problems
  - e.g., achieving goal [on(a,b) , on (b,a)]
- What's needed? A notion of *protecting* a sub-goal so it's not undone by later steps

# State-space planning

- STRIPS searches thru a space of situations (where you are, what you have, etc.)
  - Plan is a solution found by “searching” through situations to get to goal
- **Progression planners** search forward from initial state to goal state
  - Usually results in a high branching factor
- **Regression planners** search backward from goal
  - OK if operators have enough information to go both ways
  - Can reduce branching: you’re only considering things relevant to goal
  - Handling a conjunction of goals is difficult (e.g., STRIPS)

# Plan-space planning

- An alternative is to **search through the space of plans**, rather than situations
- Start from a **partial plan** which is expanded and refined until a complete plan is generated
- **Refinement operators** add constraints to the partial plan and modification operators for other changes
- We can still use STRIPS-style operators:
  - Op(ACTION: RightShoe, PRECOND: RightSockOn, EFFECT: RightShoeOn)
  - Op(ACTION: RightSock, EFFECT: RightSockOn)
  - Op(ACTION: LeftShoe, PRECOND: LeftSockOn, EFFECT: LeftShoeOn)
  - Op(ACTION: LeftSock, EFFECT: leftSockOn)

could result in a partial plan of

[ ... RightShoe ... LeftShoe ... ]

# Partial-order planning

- Linear planners build plans as **totally ordered sequences** of steps
- Non-linear planners (aka partial-order planners) build plans as sets of steps with temporal constraints
  - constraints like  $S1 < S2$  if step  $S1$  must come before  $S2$
- One **refines** a partially ordered plan (POP) by either:
  - adding a new plan step, or
  - adding a new constraint to the steps already in the plan
- A POP can be **linearized** (converted to a totally ordered plan) by topological sorting

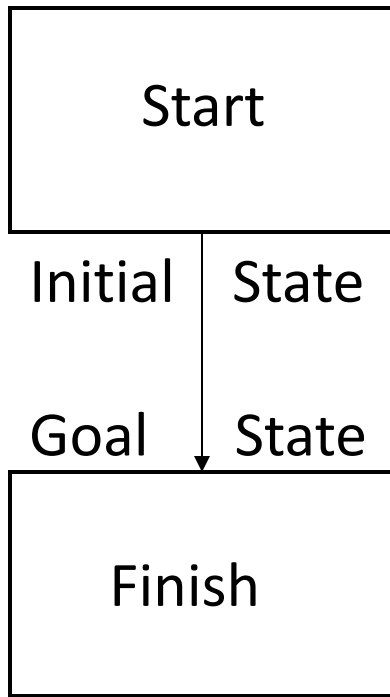


# Some example domains

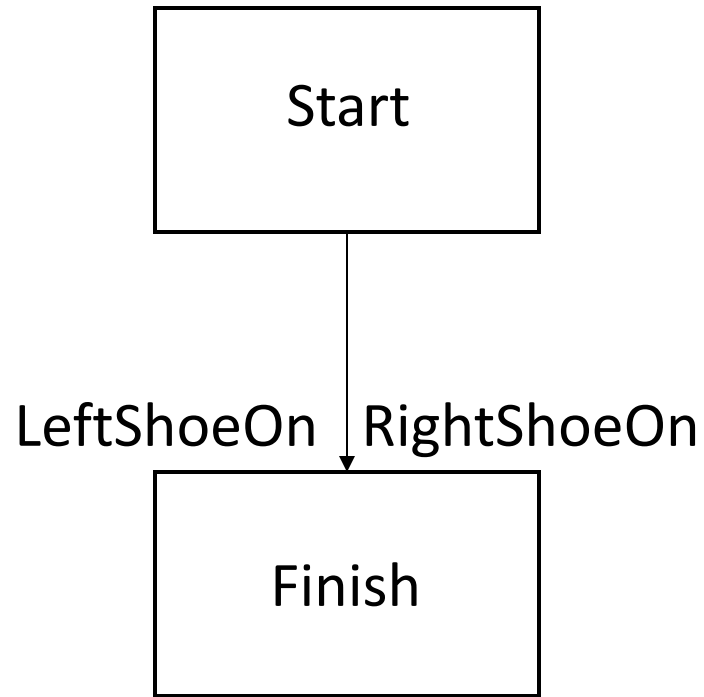
We'll use some simple problems with a real world flavor to illustrate planning problems and algorithms

- Putting on your socks and shoes in the morning
  - Actions like put-on-left-sock, put-on-right-shoe
- Planning a shopping trip involving buying several kinds of items
  - Actions like go(X), buy(Y)

# A simple graphical notation



(a)



(b)



# Least commitment

- Non-linear planners embody the principle of **least commitment**
  - only choose actions, orderings & variable bindings absolutely necessary, postponing other decisions
  - avoids early commitment to decisions that don't really matter
- Linear planners always choose to add a plan step in a particular place in the sequence
- Non-linear planners choose to add a step and possibly some temporal constraints

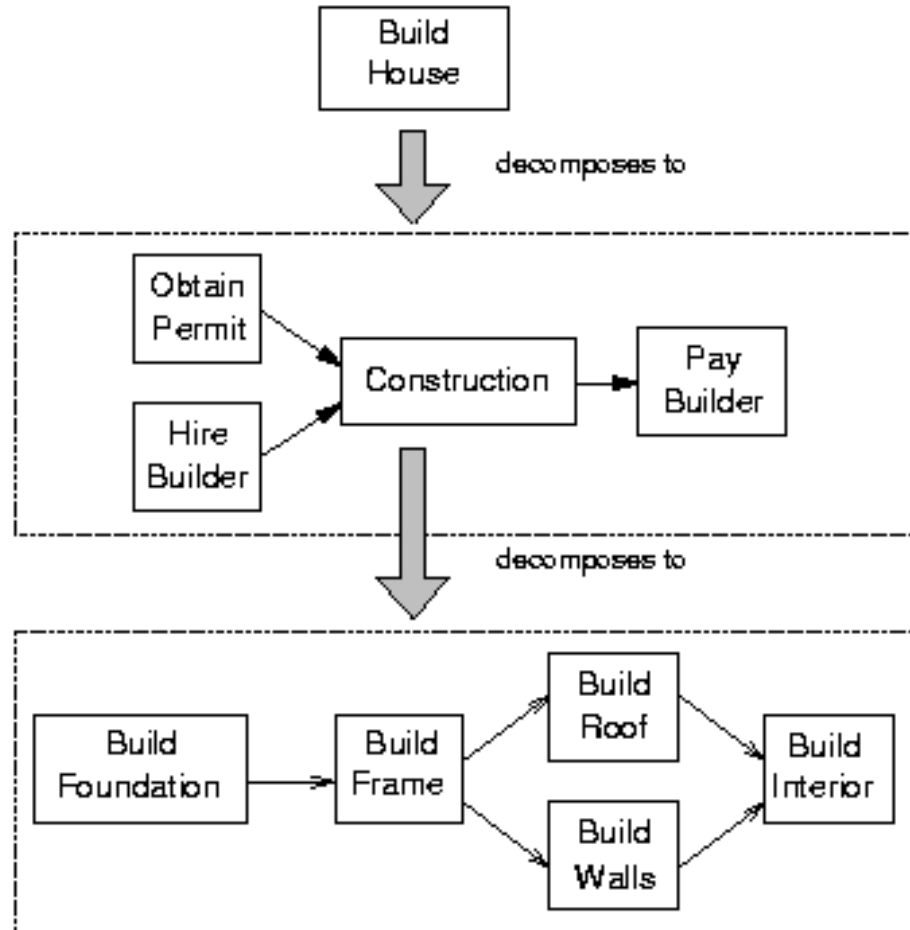
# Real-world planning domains

- Real-world domains are complex and don't satisfy assumptions of STRIPS or methods
  - Some of the characteristics we may need to handle:
    - Modeling and reasoning about resources
    - Representing and reasoning about time
    - Planning at different levels of abstractions
    - Conditional outcomes of actions
    - Uncertain outcomes of actions
    - Exogenous events
    - Incremental plan development
    - Dynamic real-time re-planning
- } Scheduling
- } Planning under uncertainty
- } HTN planning

# Hierarchical decomposition

- Hierarchical decomposition, or hierarchical task network (HTN) planning, uses **abstract operators** to **incrementally** decompose a planning problem from a **high-level goal** statement to a **primitive plan network**
- **Primitive operators** represent actions that are **executable**, and can appear in the final plan
- **Non-primitive operators** represent **goals** (equivalently, **abstract actions**) that require further decomposition (or operationalization) to be executed
- There is no “right” set of primitive actions: One agent’s goals are another agent’s actions!

# HTN planning: example



# HTN operator: Example

OPERATOR decompose

PURPOSE: Construction

CONSTRAINTS:

Length (Frame)  $\leq$  Length (Foundation),

Strength (Foundation)  $>$  Wt(Frame) + Wt(Roof)

+ Wt(Walls) + Wt(Interior) + Wt(Contents)

PLOT: Build (Foundation)

Build (Frame)

PARALLEL

Build (Roof)

Build (Walls)

END PARALLEL

Build (Interior)



# Planning summary

- Planning representations
  - Situation calculus
  - STRIPS representation: Preconditions and effects
- Planning approaches
  - State-space search (STRIPS, forward chaining, ....)
  - Plan-space search (partial-order planning, HTN, ...)
  - Constraint-based search (GraphPlan, SATplan, ...)
- Search strategies
  - Forward planning
  - Goal regression
  - Backward planning
  - Least-commitment
  - Nonlinear planning