Chapter 10 Classical Planning

CS4811 – Artificial Intelligence

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Outline

Planning systems

PDDL (Planning Domain Definition Language)

Planning algorithms

- Forward chaining
- Backward chaining
- Partial-order planning

Applications

Motivating reasons

- Planning is a component of intelligent behavior.
- It has lots of applications.

What is planning?

- A ***planner*** is a system that finds a sequence of actions to accomplish a specific task
- A planner synthesizes a plan

What is planning? (cont'd)

- **The main components of a *planning problem* are:**
  - **a description of the starting situation (*the initial state*), (the world now)**
  - **a description of the desired situation (*the goal state*), (how should the world be)**
  - **the actions available to the executing agent (*operator library*, a.k.a. *domain theory*) (possible actions to change the world)**
- **Formally, a *classical planning problem* is a triple:**

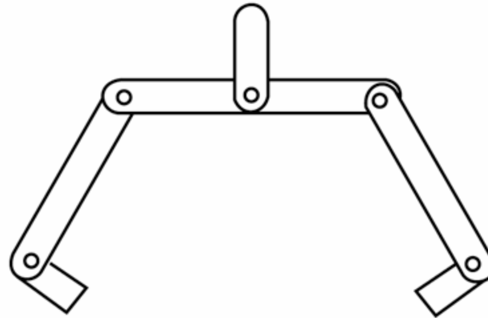
<I, G, D>, where

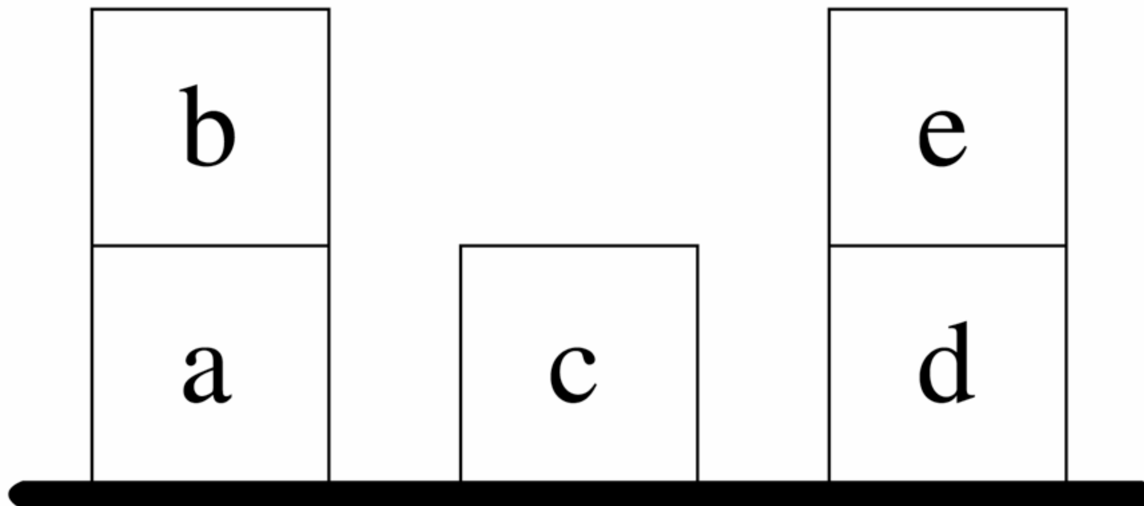
I is the initial state,
G is the goal state, and
D is the domain theory.

Characteristics of *classical planners*

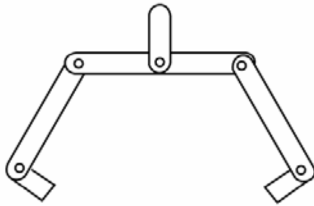
- **They operate on basic STRIPS actions.**
- **Important assumptions:**
  - **the agent is the only source of change in the world, otherwise the environment is static.**
  - **all the actions are deterministic.**
  - **the agent is omniscient: knows everything it needs to know about start state and effects of actions.**
  - **the goals are categorical, the plan is considered successful iff all the goals are achieved.**

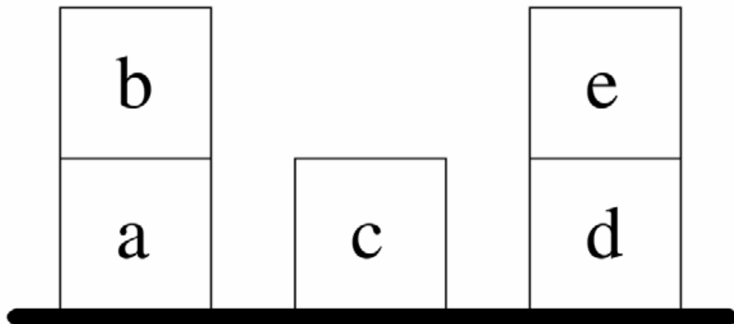
The blocks world





Represent this world using predicates





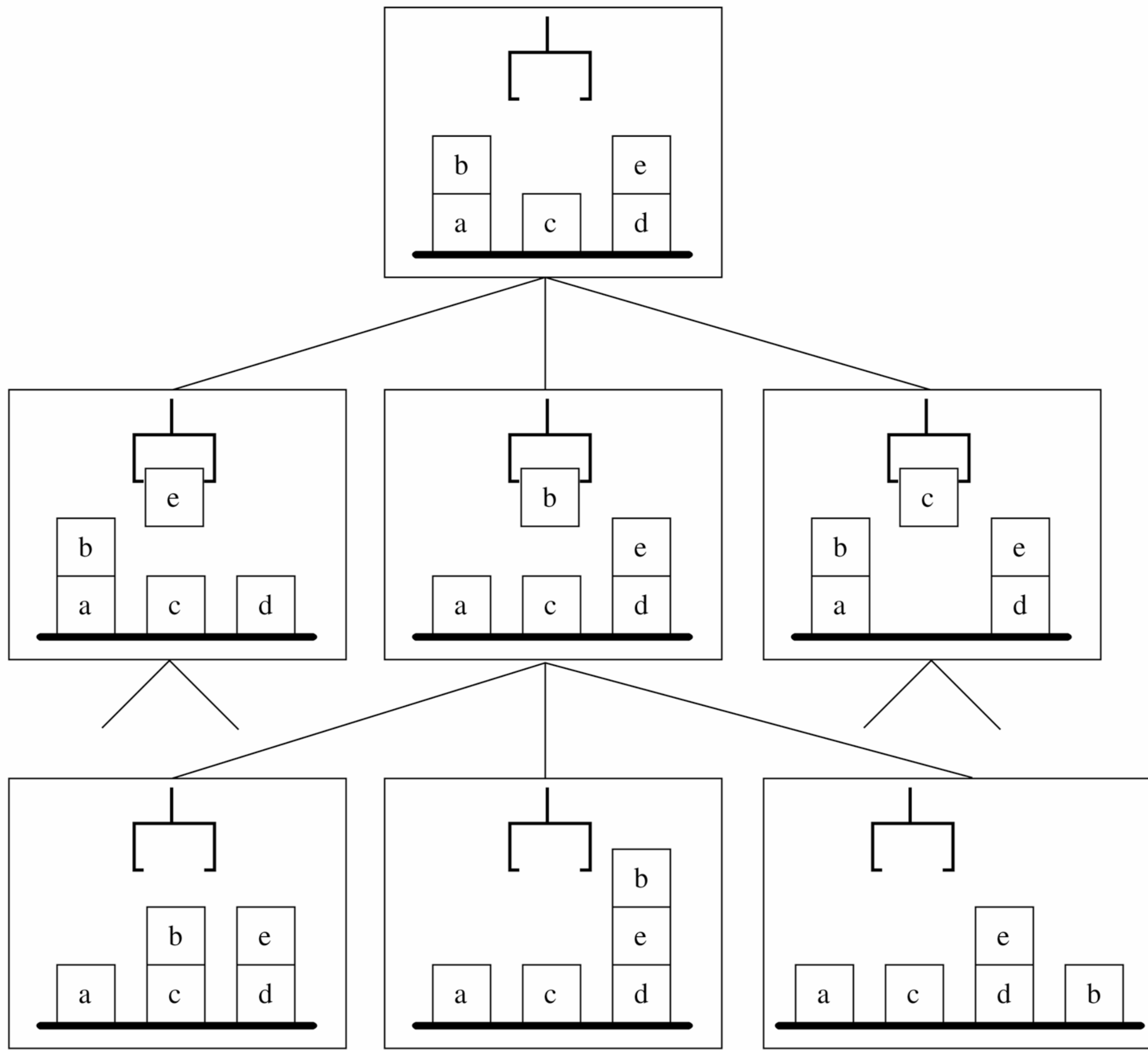
ontable(a)
ontable(c)
ontable(d)
on(b,a)
on(e,d)
clear(b)
clear(c)
clear(e)
gripping()

The robot arm can perform these tasks

- **pickup (W)**: pick up block W from its current location on the table and hold it
- **putdown (W)**: place block W on the table
- **stack (U, V)**: place block U on top of block V
- **unstack (U, V)**: remove block U from the top of block V and hold it

All assume that the robot arm can precisely reach the block.

Portion of the search space of the blocks world example



The STRIPS representation

Special purpose representation.

An operator is defined in terms of its:

name,
parameters,
preconditions, and
results.

A planner is a special purpose algorithm, i.e., it’s not a general purpose logic theorem prover.

Four operators for the blocks world

pickup(X)	P: gripping() $\wedge$ clear(X) $\wedge$ ontable(X) A: gripping(X) D: ontable(X) $\wedge$ gripping()
putdown(X)	P: gripping(X) A: ontable(X) $\wedge$ gripping() $\wedge$ clear(X) D: gripping(X)
stack(X,Y)	P: gripping(X) $\wedge$ clear(Y) A: on(X,Y) $\wedge$ gripping() $\wedge$ clear(X) D: gripping(X) $\wedge$ clear(Y)
unstack(X,Y)	P: gripping() $\wedge$ clear(X) $\wedge$ on(X,Y) A: gripping(X) $\wedge$ clear(Y) D: on(X,Y) $\wedge$ gripping()

Notice the simplification

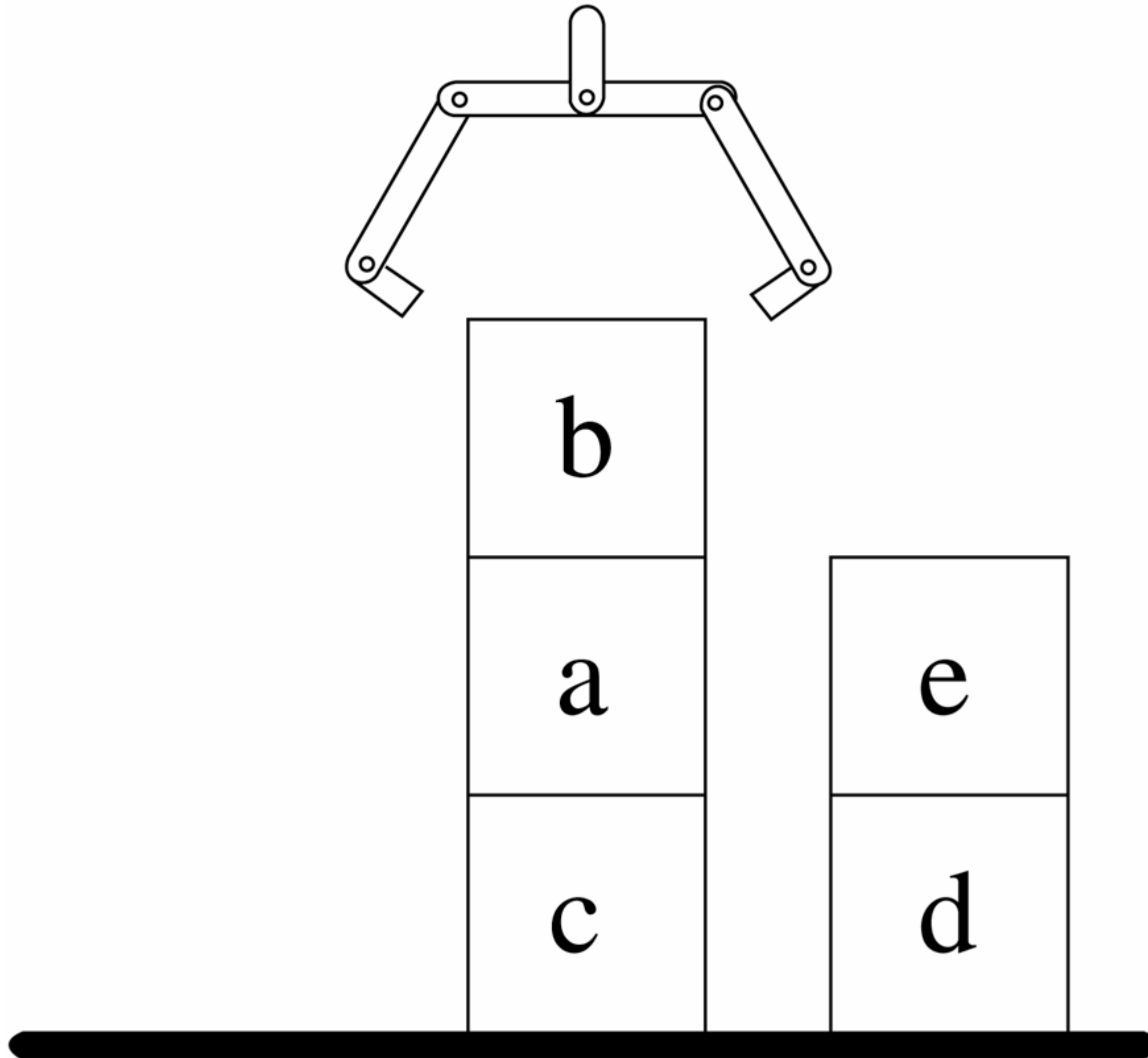
***Preconditions*, *add lists*, and *delete lists* are all conjunctions. We don’t have the full power of predicate logic.**

The same applies to *goals*. Goals are conjunctions of predicates.

A detail:

Why do we have two operators for picking up (pickup and unstack), and two for putting down (putdown and stack)?

A goal state for the blocks world



A state space algorithm for STRIPS operators

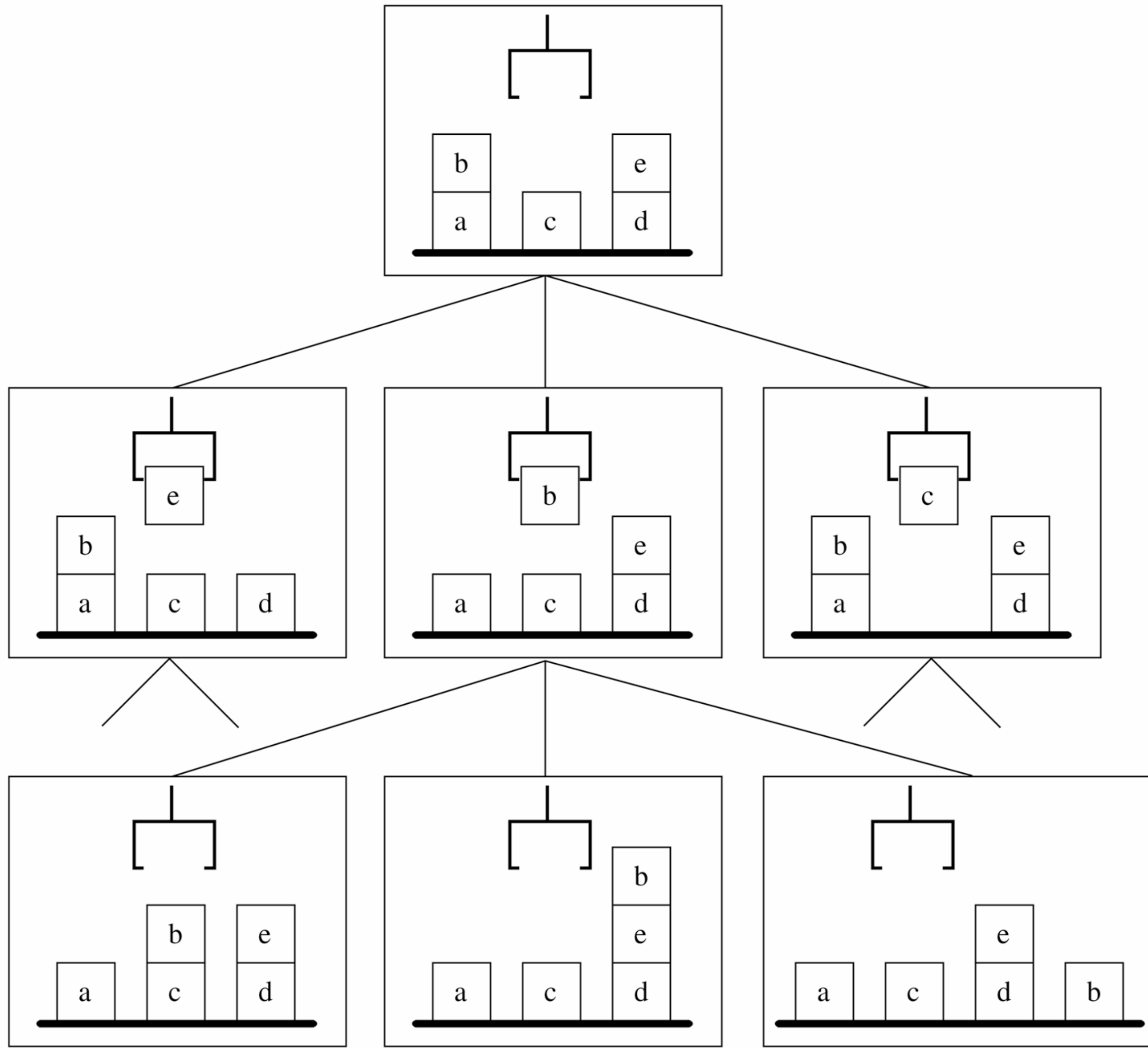
Search the space of situations (or states). This means each node in the search tree is a state.

The root of the tree is the start state.

Operators are the means of transition from each node to its children.

The goal test involves seeing if the set of goals is a subset of the current situation.

Now, the following graph makes much more sense



Problems in representation

***Frame problem*: List everything that does not change. It no more is a significant problem because what is not listed as changing (via the add and delete lists) is assumed to be not changing.**

***Qualification problem*: Can we list every precondition for an action? For instance, in order for PICKUP to work, the block should not be glued to the table, it should not be nailed to the table, …**

It still is a problem. A partial solution is to prioritize preconditions, i.e., separate out the preconditions that are worth achieving.

Problems in representation (cont'd)

***Ramification problem*: Can we list every result of an action? For instance, if a block is picked up its shadow changes location, the weight on the table decreases, ...**

It still is a problem. A partial solution is to code rules so that inferences can be made. For instance, allow rules to calculate where the shadow would be, given the positions of the light source and the object. When the position of the object changes, its shadow changes too.

The gripper domain

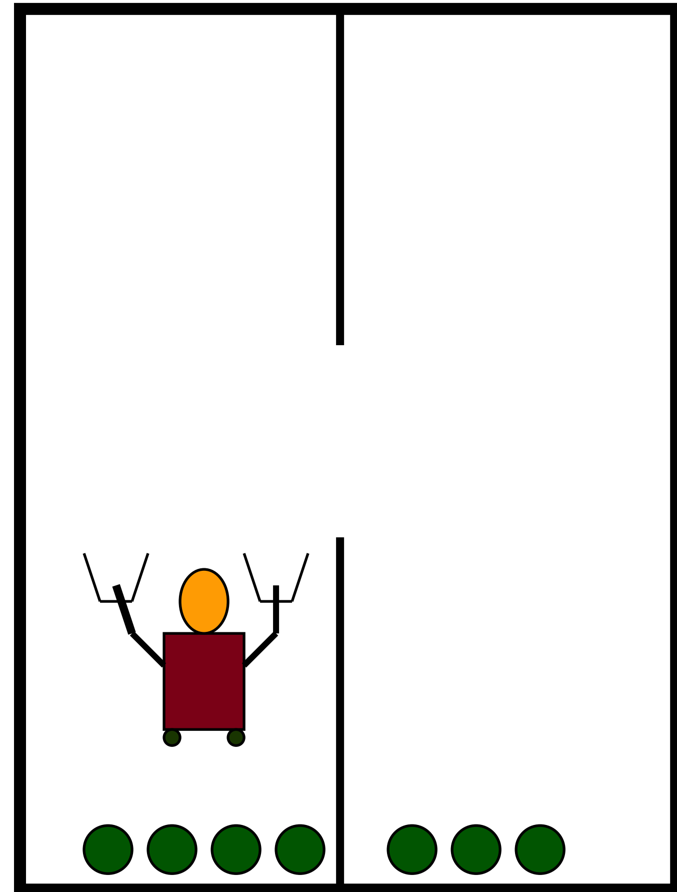
The agent is a robot with two grippers (left and right)

There are two rooms (rooma and roomb)

There are a number of balls in each room

Operators:

- PICK
- DROP
- MOVE



A “deterministic” plan

Pick ball1 rooma right

Move rooma roomb

Drop ball1 roomb right

Remember: the plans are generated “offline,” no observability, nothing can go wrong.

The gripper domain is interesting because parallelism is possible: can pick with both grippers at the same time.

How to define a planning problem

- **Create a *domain file:* contains the domain behavior, simply the operators**
- **Create a *problem file:* contains the initial state and the goal**

The *domain definition* for the gripper domain

```
(define (domain gripper-strips)
  (:predicates   (room ?r)         (ball ?b)
                 (gripper ?g)      (at-robby ?r)
                 (at ?b ?r)        (free ?g)
                 (carry ?o ?g))
```

name of the domain

"?" indicates a variable

name of the action

```
  (:action move
       :parameters (?from ?to)
       :precondition (and (room ?from) (room ?to)
                          (at-robby ?from))
       :effect (and (at-robby ?to)
                    (not (at-robby ?from))))
```

combined add and delete lists

The *domain definition* for the gripper domain (cont'd)

```
(:action pick
    :parameters (?obj ?room ?gripper)
    :precondition (and (ball ?obj) (room ?room)
      (gripper ?gripper) (at ?obj ?room)
      (at-robby ?room) (free ?gripper))
    :effect (and (carry ?obj ?gripper)
      (not (at ?obj ?room)) (not (free ?gripper))))
```

The *domain definition* for the gripper domain (cont'd)

```
(:action drop
        :parameters (?obj ?room ?gripper)
        :precondition (and (ball ?obj) (room ?room)
           (gripper ?gripper) (at-robby ?room)
           (carrying ?obj ?gripper))
        :effect (and (at ?obj ?room) (free ?gripper)
           (not (carry ?obj ?gripper))))))
```

An example *problem definition* for the gripper domain

```
(define (problem strips-gripper2)
  (:domain gripper-strips)

  (:objects rooma roomb ball1 ball2 left right)

  (:init (room rooma)        (room roomb)
         (ball ball1)        (ball ball2)
         (gripper left)      (gripper right)

         (at-robby rooma)
         (free left)         (free right)
         (at ball1 rooma)    (at ball2 rooma) )

  (:goal (at ball1 roomb)))
```

Running VHPOP

Once the domain and problem definitions are in files gripper-domain.pddl and gripper-2.pddl respectively, the following command runs Vhpop:

```
vhpop gripper-domain.pddl gripper-2.pddl
```

The output will be:

```
;strips-gripper2
1:(pick ball1 rooma right)
2:(move rooma roomb)
3:(drop ball1 roomb right)
Time: 0 msec.
```

“pddl” is the extension for the planning domain definition language.

Why is planning a hard problem?

It is due to the large branching factor and the overwhelming number of possibilities.

There is usually no way to separate out the relevant operators. Take the previous example, and imagine that there are 100 balls, just two rooms, and two grippers. Again, the goal is to take 1 ball to the other room.

How many PICK operators are possible in the initial situation?

pick
    :parameters (?obj ?room ?gripper)

That is only one part of the branching factor, the robot could also move without picking up anything.

Why is planning a hard problem? (cont'd)

Also, goal interactions is a major problem. In planning, goal-directed search seems to make much more sense, but unfortunately cannot address the exponential explosion. This time, the branching factor increases due to the many ways of resolving the interactions.

When subgoals are compatible, i.e., they do not interact, they are said to be *linear* (or *independent*, or *serializable*).

Life is easier for a planner when the subgoals are independent because then divide-and-conquer works.

How to deal with the exponential explosion?

Use goal-directed algorithms

Use domain-independent heuristics

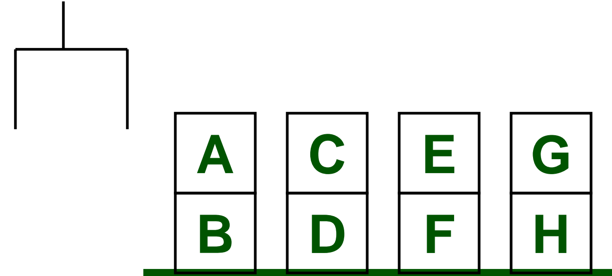
Use domain-dependent heuristics
(we need a language to specify them)

A sampler of planning algorithms

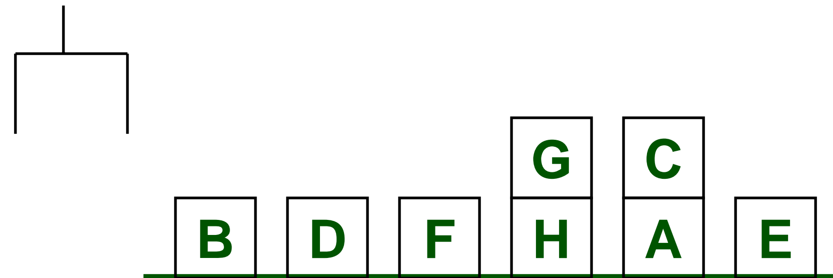
- **Forward chaining**
  - Work in a state space
  - Start with the initial state, try to reach the goal state using forward progression
- **Backward chaining**
  - Work in a state space
  - Start with the goal state, try to reach the initial state using backward regression
- **Partial order planning**
  - Work in a plan space
  - Start with an empty plan, work from the goal to reach a complete plan

Forward chaining

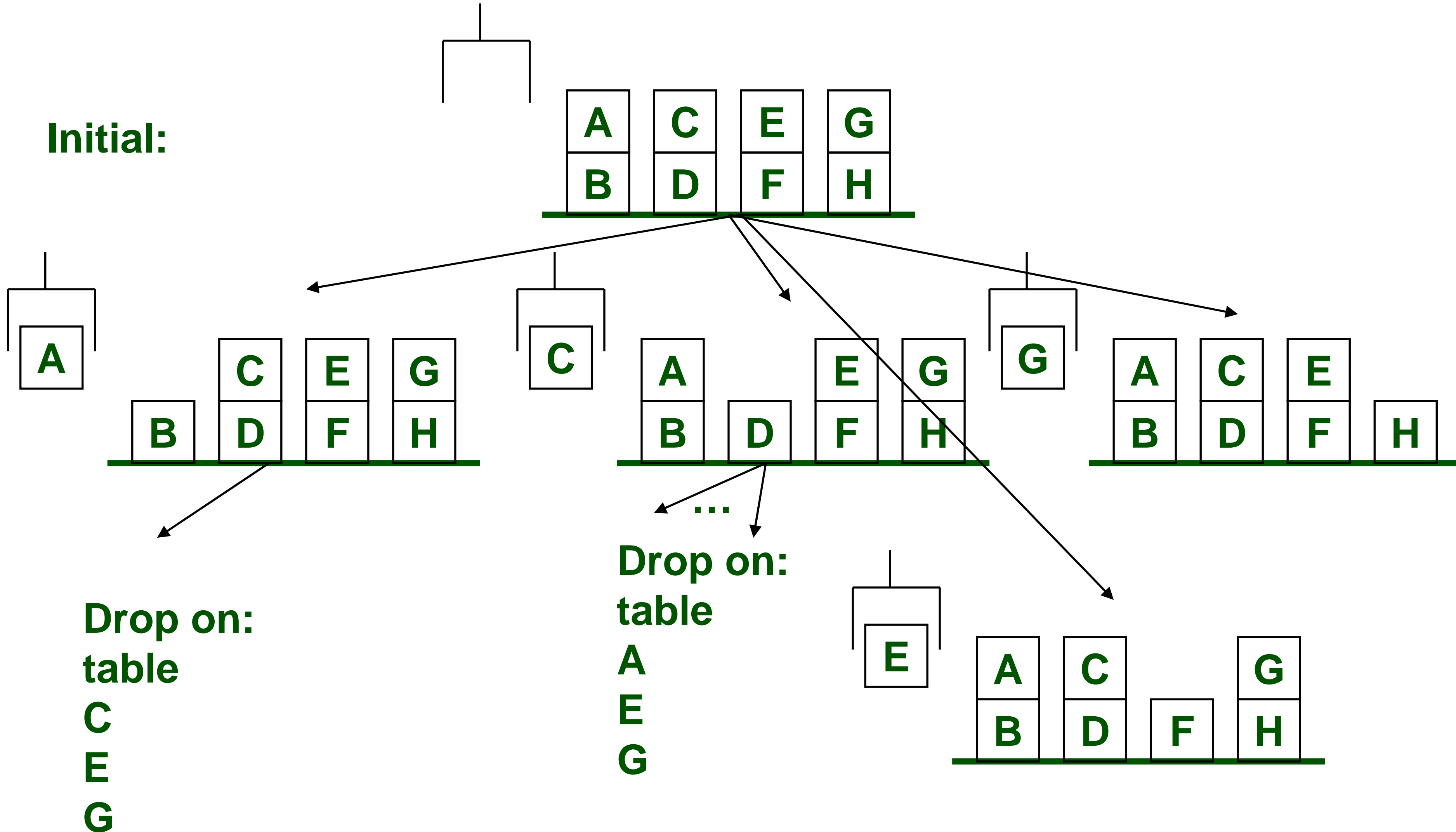
Initial:



Goal :



1st and 2nd levels of search

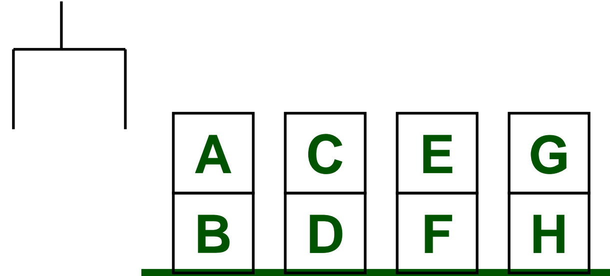


Results

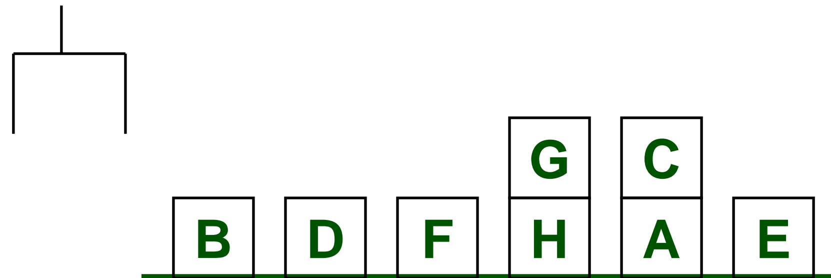
- **A plan is:**
  - **unstack (A, B)**
  - **putdown (A)**
  - **unstack (C, D)**
  - **stack (C, A)**
  - **unstack (E, F)**
  - **putdown (F)**
- **Notice that the final locations of D, F, G, and H need not be specified**
- **Also notice that D, F, G, and H will never need to be moved. But there are states in the search space which are a result of moving these. Working backwards from the goal might help.**

Backward chaining

Initial:



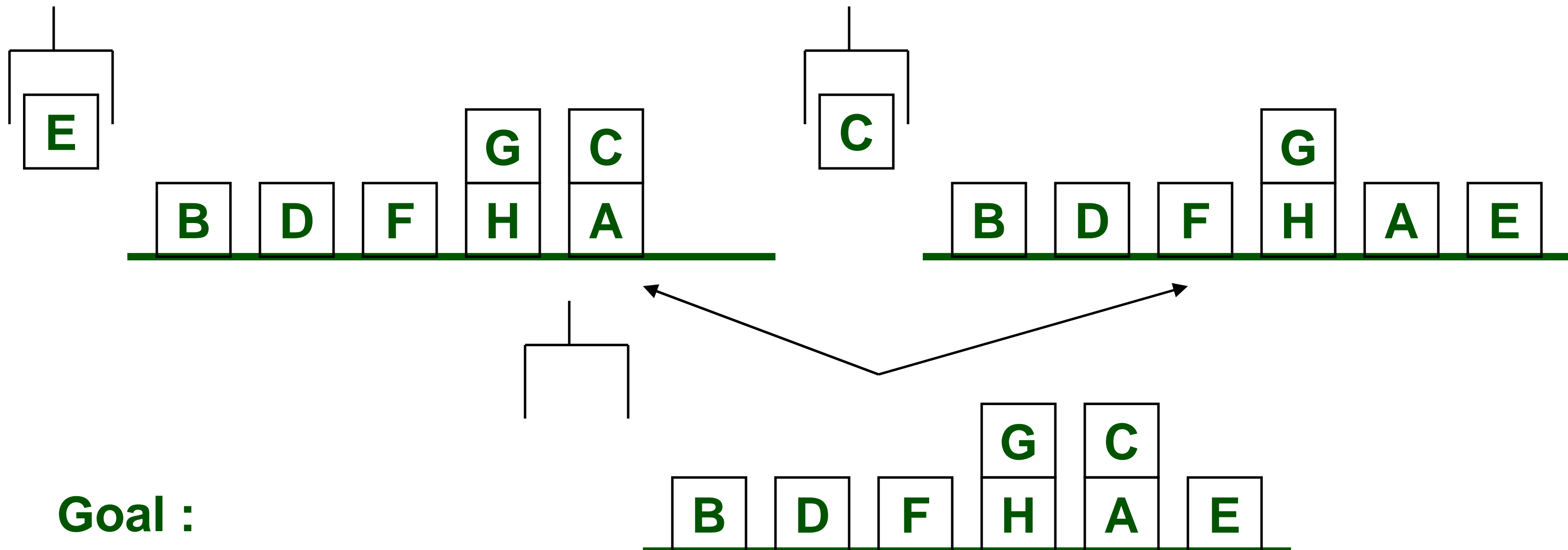
Goal :



1st level of search

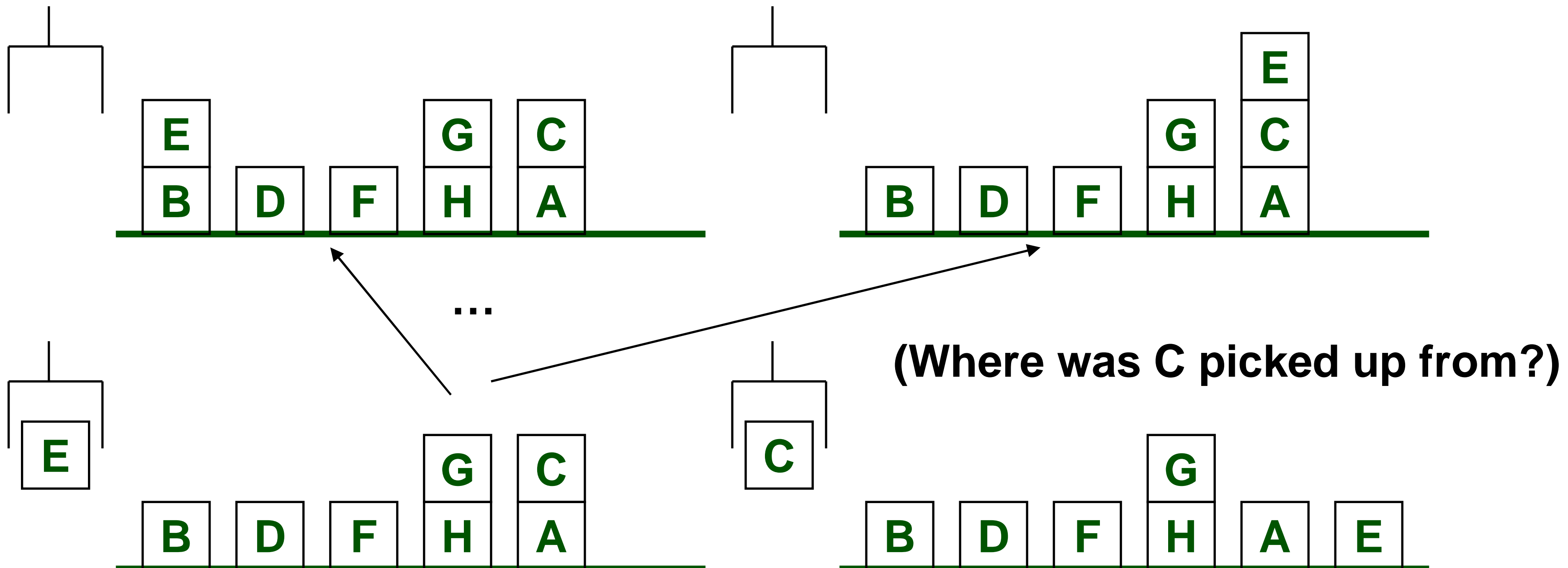
For E to be on the table, the last action must be putdown(E)

For C to be on A, the last action must be stack(C,A)



2nd level of search

Where was E picked up from?



Results

- **The same plan can be found**
  - **unstack (A, B)**
  - **putdown (A)**
  - **unstack (C, D)**
  - **stack (C, A)**
  - **unstack (E, F)**
  - **putdown (F)**
- **Now, the final locations of D, F, G, and H need to be specified**
- **Notice that D, F, G, and H will never need to be moved. But observe that from the second level on the branching factor is still high**

Partial-order planning (POP)

- Notice that the resulting plan has two parallelizable threads:

unstack (A,B) putdown (A) unstack (C,D) stack (C,A)	&	unstack (E, F) putdown (F)

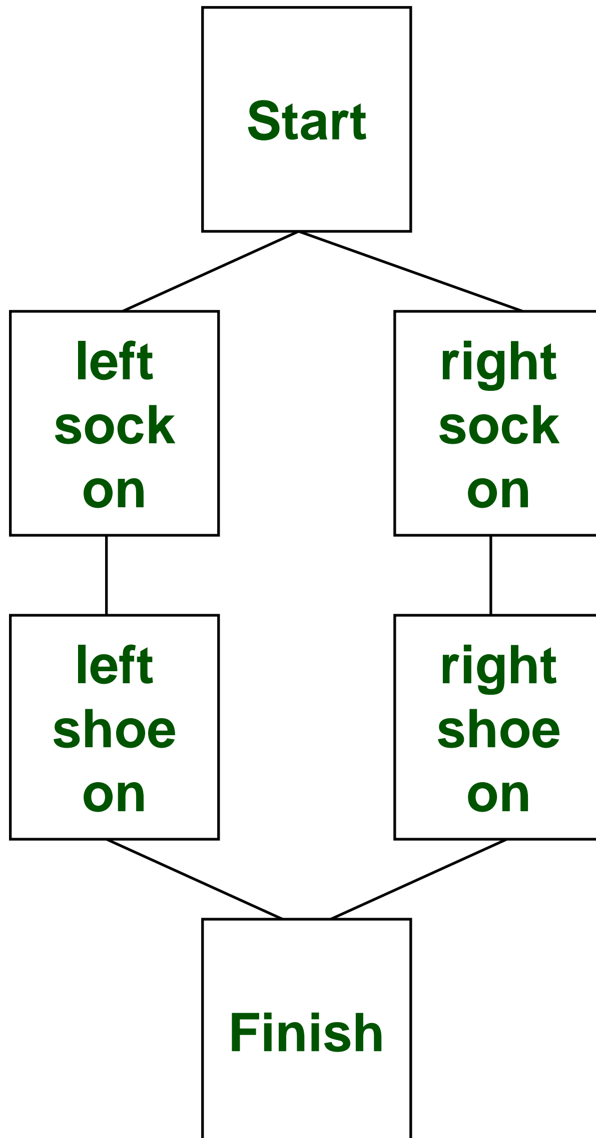
- These steps can be interleaved in 3 different ways:

unstack (E, F)	unstack (A,B)	unstack (A,B)
putdown (F)	putdown (A)	putdown (A)
unstack (A,B)	unstack (E, F)	unstack (C,D)
putdown (A)	putdown (F)	stack (C,A)
unstack (C,D)	unstack (C,D)	unstack (E, F)
stack (C,A)	stack (C,A)	putdown (F)

Partial-order planning (cont'd)

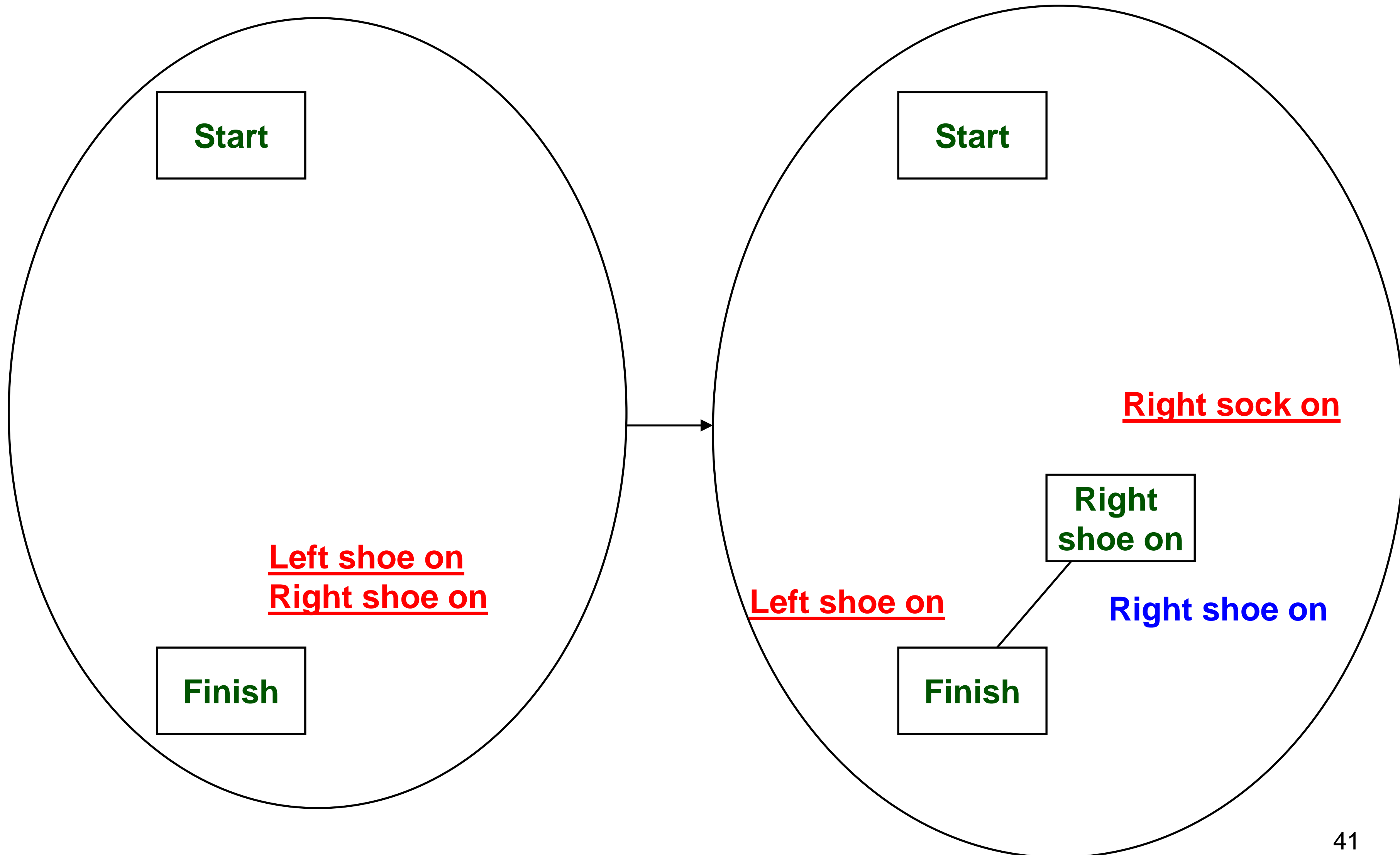
- **Idea: Do not order steps unless it is necessary**
- **Then a partially ordered plan represents several totally ordered plans**
- **That decreases the search space**
- **But still the planning problem is not solved, good heuristics are crucial**

Partial-order planning (cont'd)

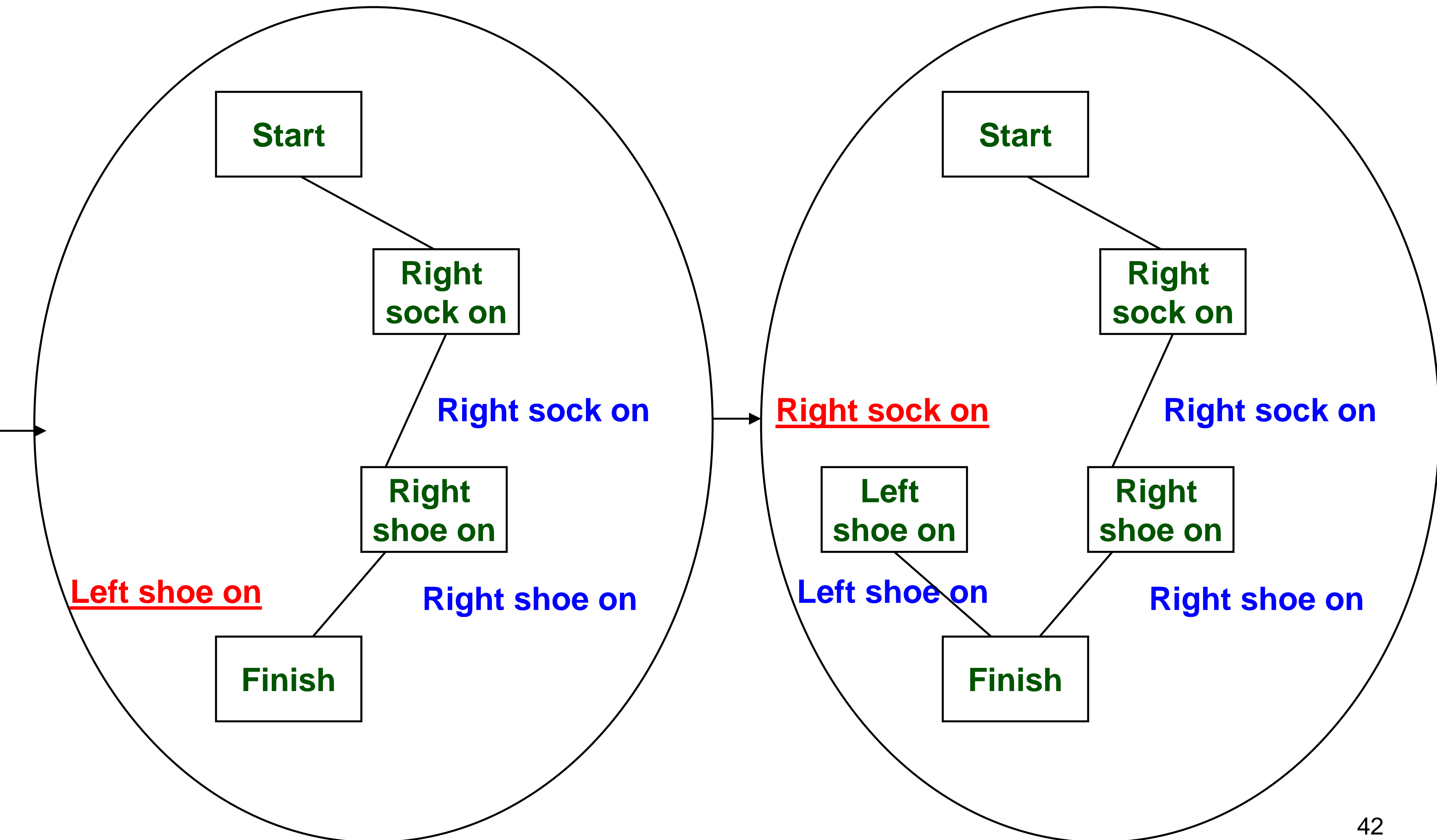


Start	Start	Start	Start	Start	Start
Left sock on	Right sock on	Left sock on	Right sock on	Left sock on	Right sock on
Left shoe on	Right shoe on	Right sock on	Left sock on	Right sock on	Left sock on
Right sock on	Left sock on	Left shoe on	Right shoe on	Right shoe on	Left shoe on
Right shoe on	Left shoe on	Right shoe on	Left shoe on	Left shoe on	Right shoe on
Finish	Finish	Finish	Finish	Finish	Finish

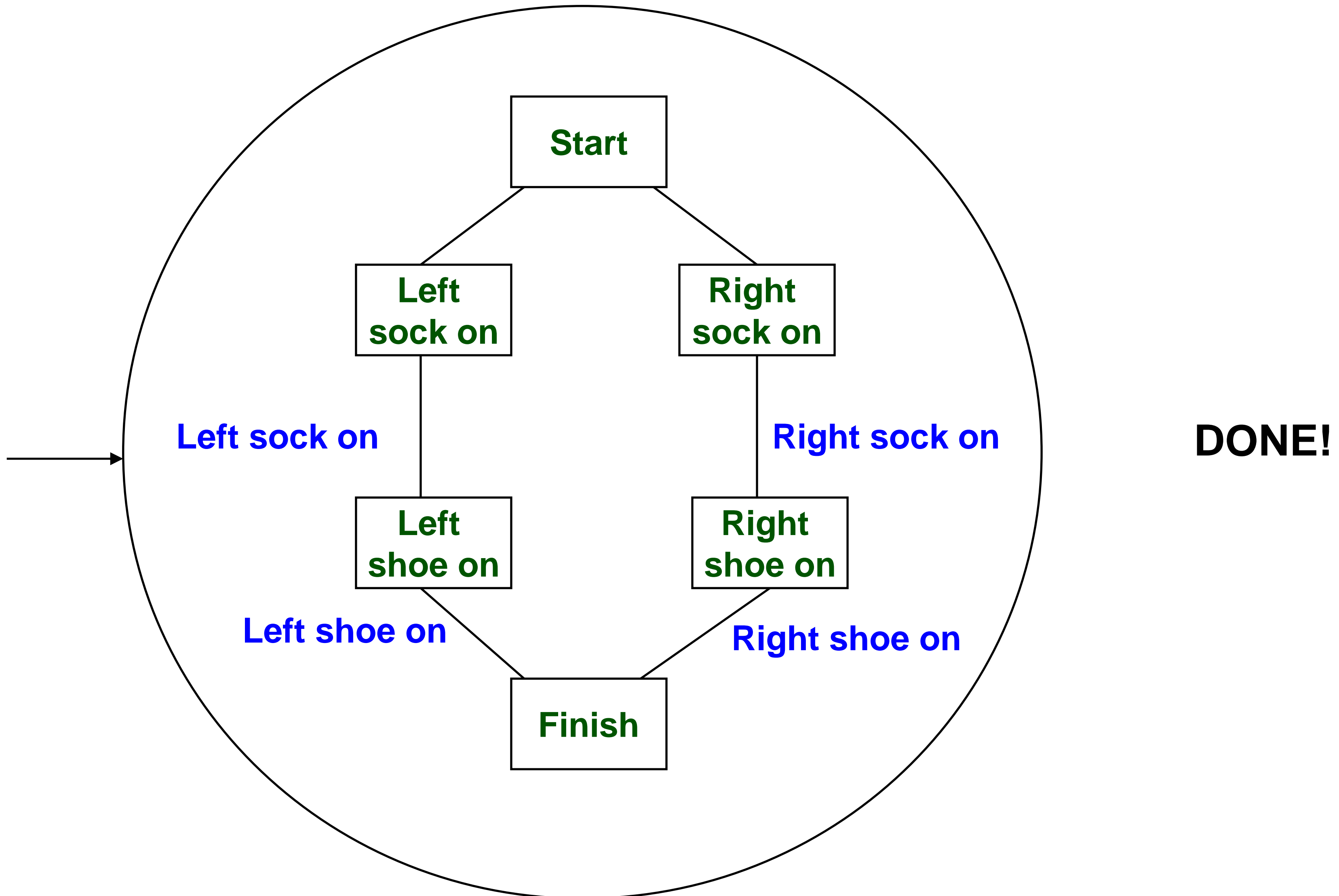
POP plan generation



POP plan generation (cont'd)



POP plan generation (cont'd)



DONE!

Comments on partial order planning

- The previous plan was generated in a straightforward manner but usually extensive search is needed.
- In the previous example there was always just one plan in the search space, normally there will be many (see the GRIPPER results).
- There is no explicit notion of a state.

Sample runs with VHPOP

- Ran increasingly larger gripper problems on wopr.
- S+OC is the older heuristic: the estimated number of steps to complete the plan is number of steps + number of open conditions.
- ADD uses a ***plan graph*** to estimate the “distance” to a complete plan.
- Both heuristics are domain independent.

Sample runs with VHPOP (cont'd)

In the examples/ directory

```
../vhpop –f static –h S+OC gripper-domain.pddl gripper-2.pddl
../vhpop –f static –h ADD   gripper-domain.pddl gripper-2.pddl
```

Run times in milliseconds

Gripper Problem	Number of Steps	S+OC heuristic	ADD heuristic
2	3	2	13
4	9	193	109
6	15	79734	562
8	21	> 10 min	1937
10	27	---	4691
12	33	---	17250
20	59	---	326718

Applications of planning

- **Robotics**
  - **Shakey, the robot at SRI was the initial motivator.**
  - **However, several other techniques are used for path-planning etc.**
  - **Most robotic systems are *reactive.***

- **Games**
The story is a plan and a different one can be constructed for each game.

- **Web applications**
Formulating query plans, using web services.

- **Crisis response**
Oil spill, forest fire, emergency evacuation.

Applications of planning (cont'd)

- **Space**
  Autonomous spacecraft, self-healing systems.
- **Device control**
  Elevator control, control software for modular devices.
- **Military planning.**
- **And many others.**

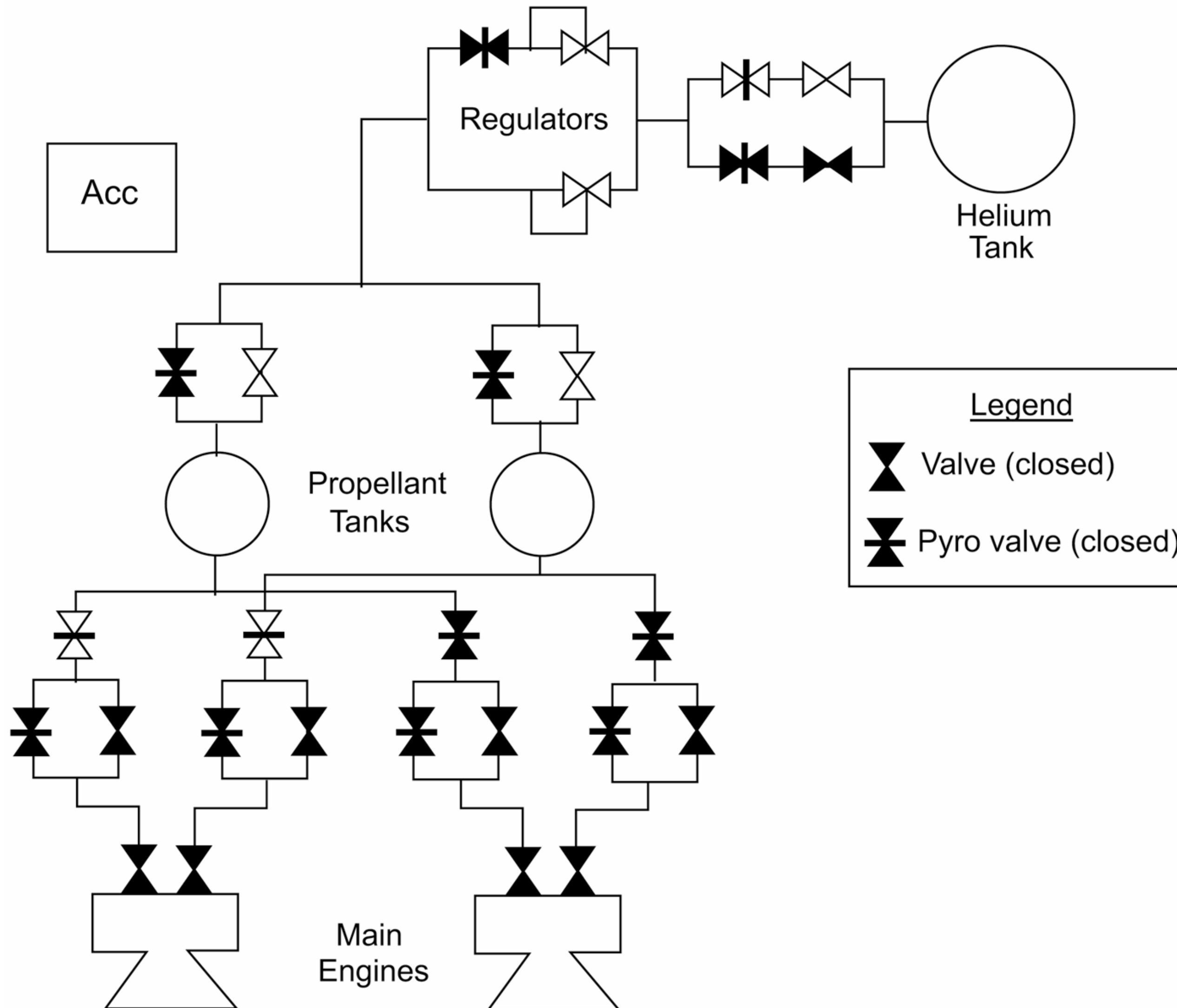
Model-based reactive configuration management (Williams and Nayak, 1996a)

Intelligent space probes that autonomously explore the solar system.

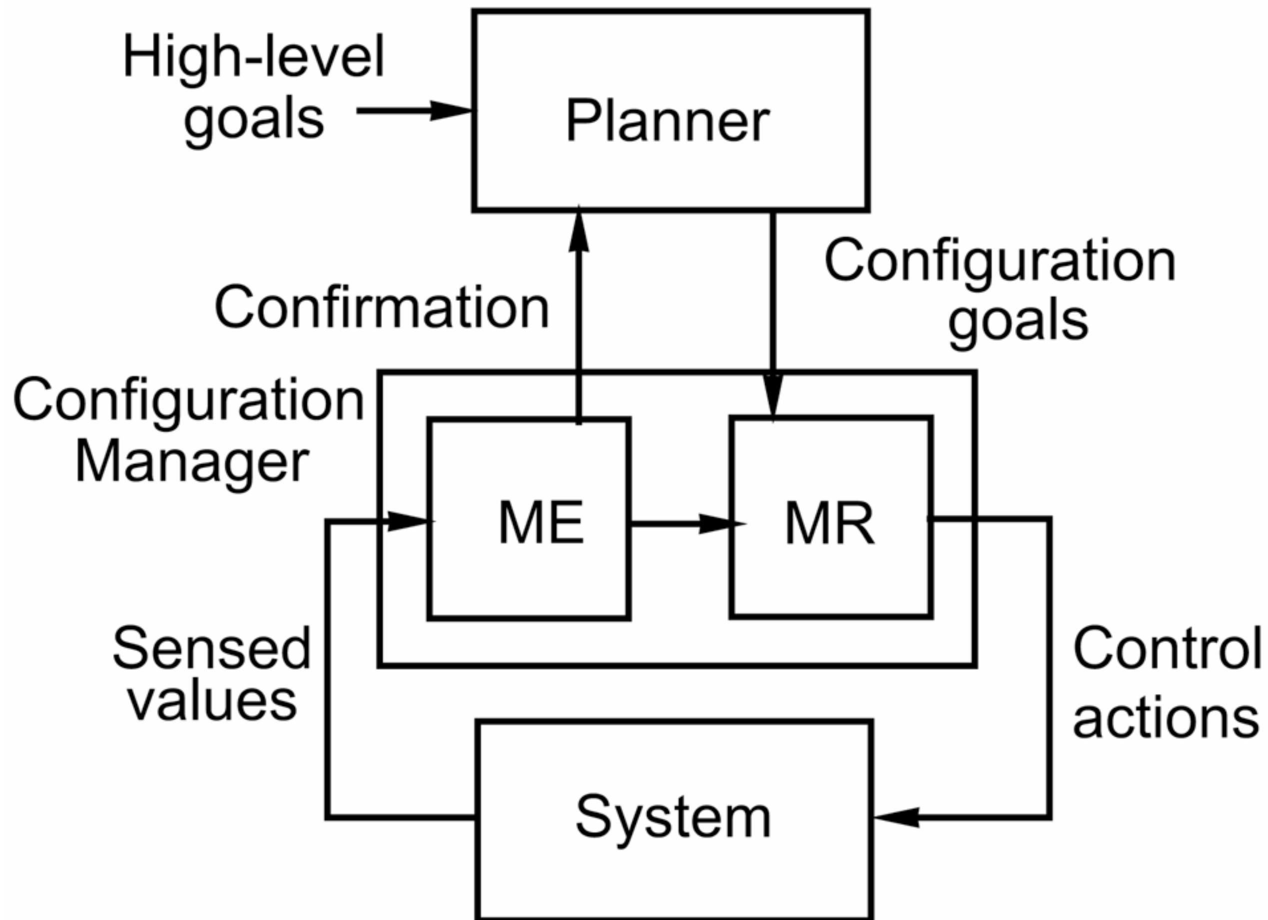
The spacecraft needs to:

- **radically reconfigure its control regime in response to failures,**
- **plan around these failures during its remaining flight.**

A schematic of the simplified Livingstone propulsion system (Williams and Nayak ,1996)

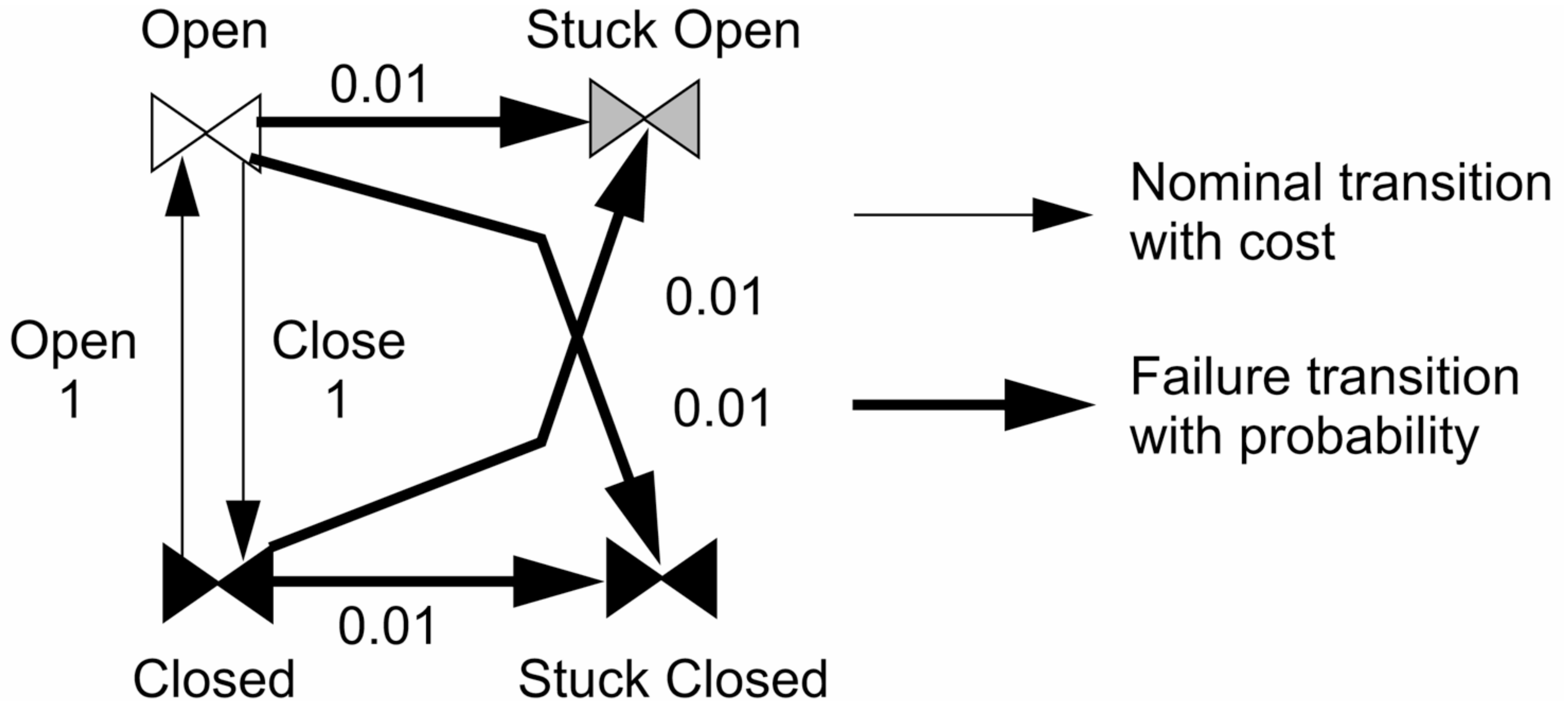


A model-based configuration management system (Williams and Nayak, 1996)

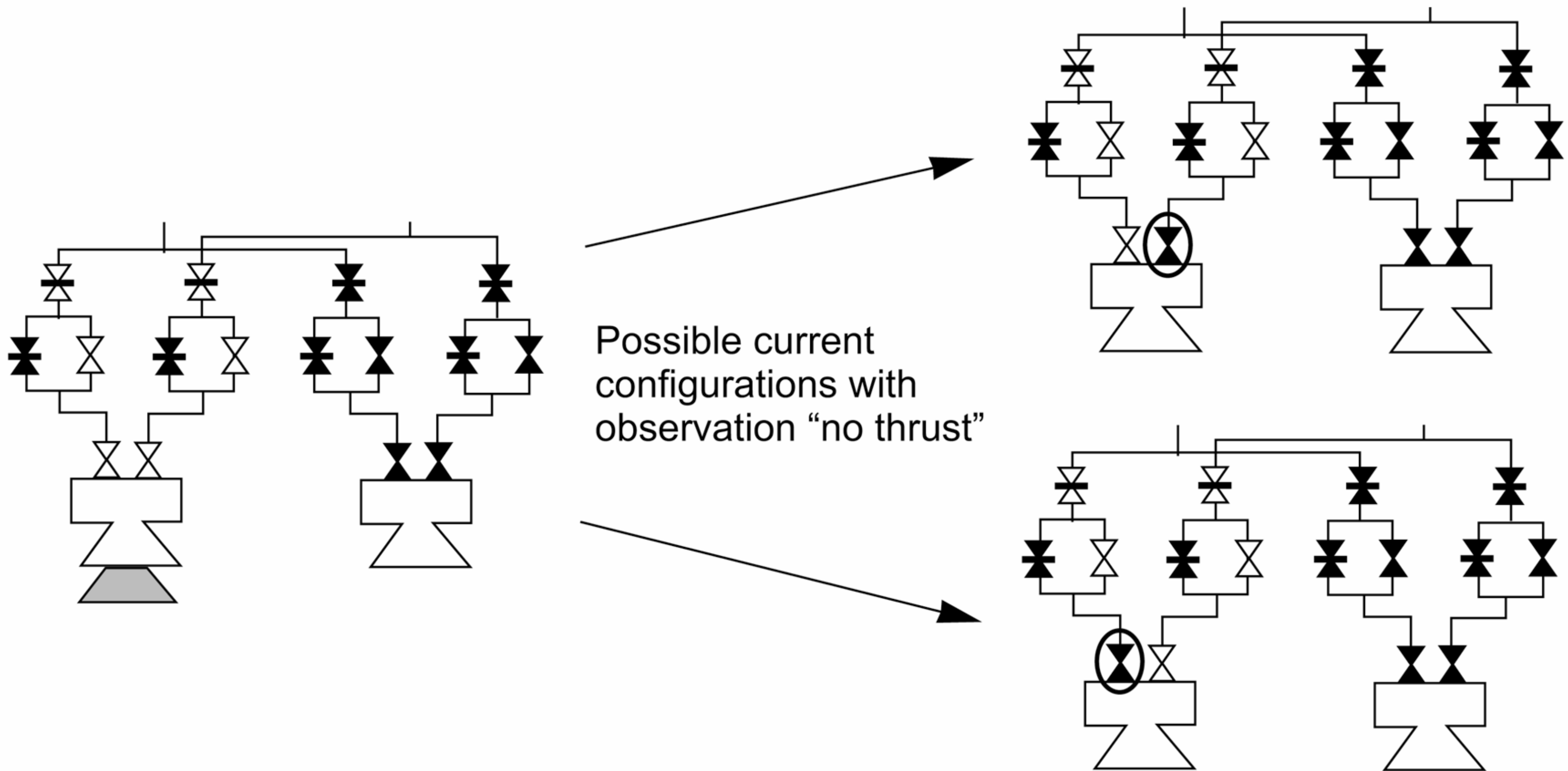


ME: mode estimation **MR: mode reconfiguration**

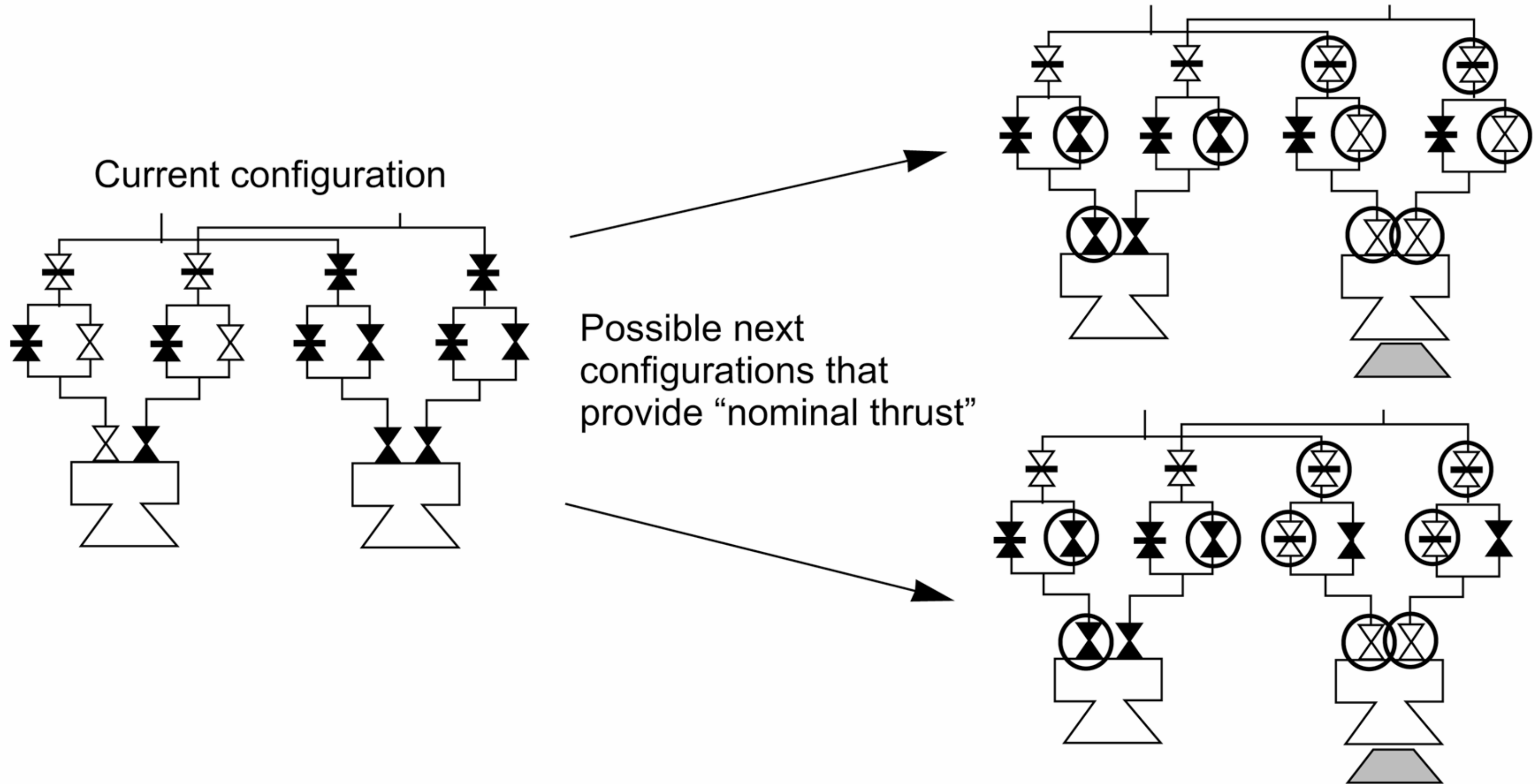
The transition system model of a valve (Williams and Nayak, 1996a)



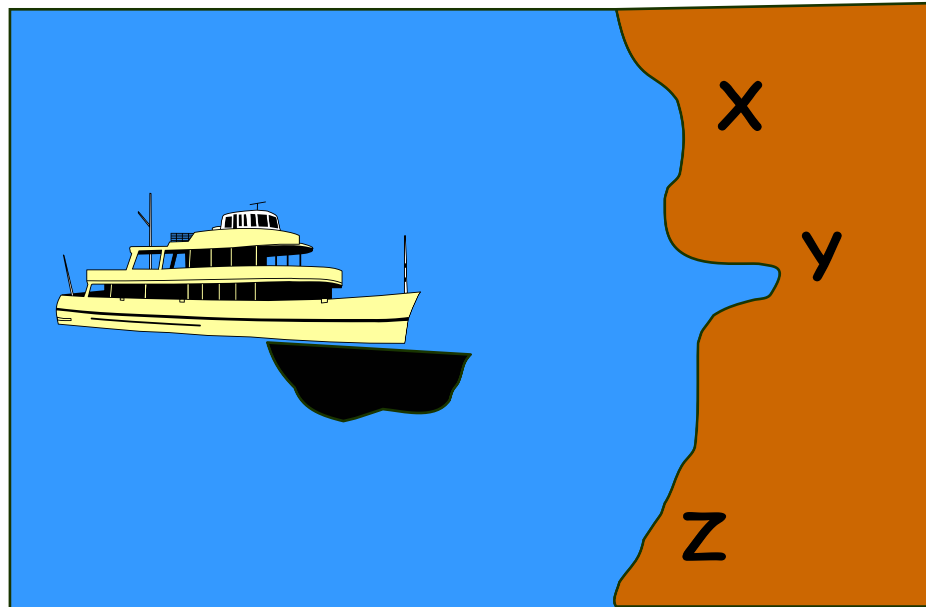
Mode estimation (Williams and Nayak, 1996a)



Mode reconfiguration (MR)
(Williams and Nayak, 1996a)



Oil spill response planning

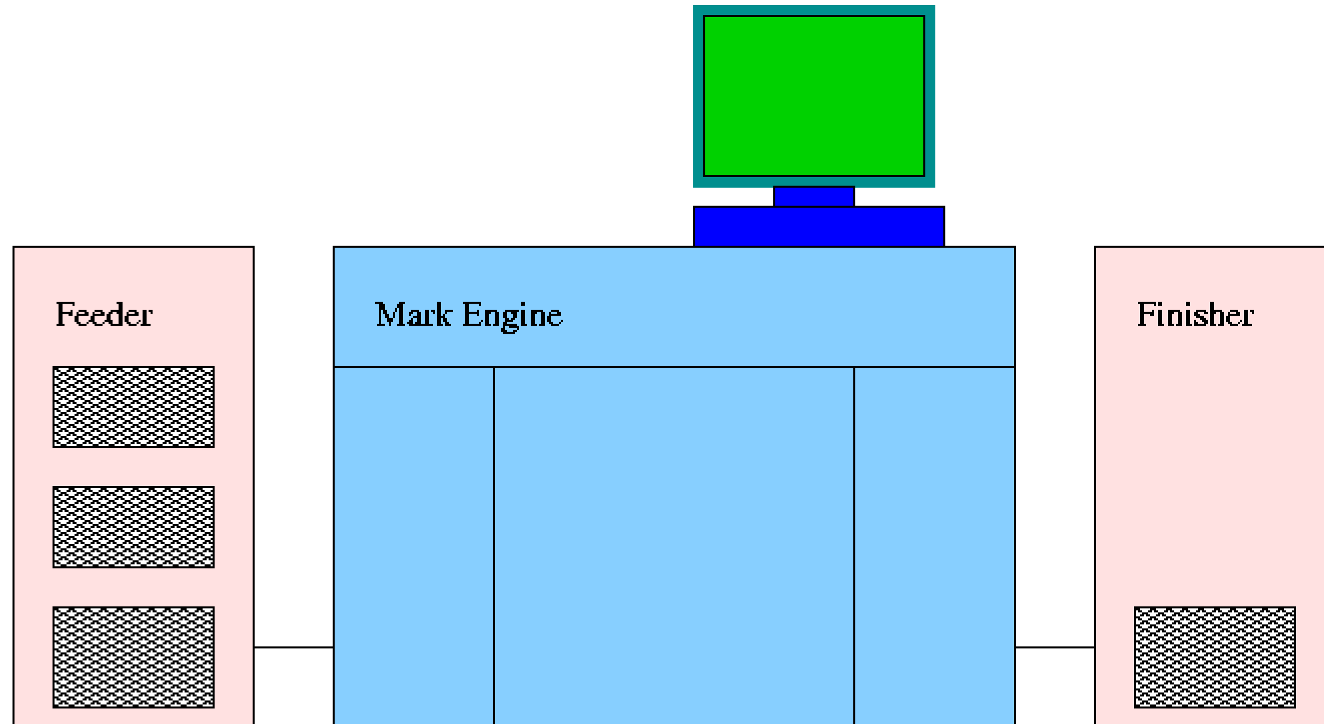


(Desimone & Agosto 1994)

Main goals: stabilize discharge, clean water, protect sensitive shore areas

The objective was to estimate the equipment required rather than to execute the plan

A modern photocopier

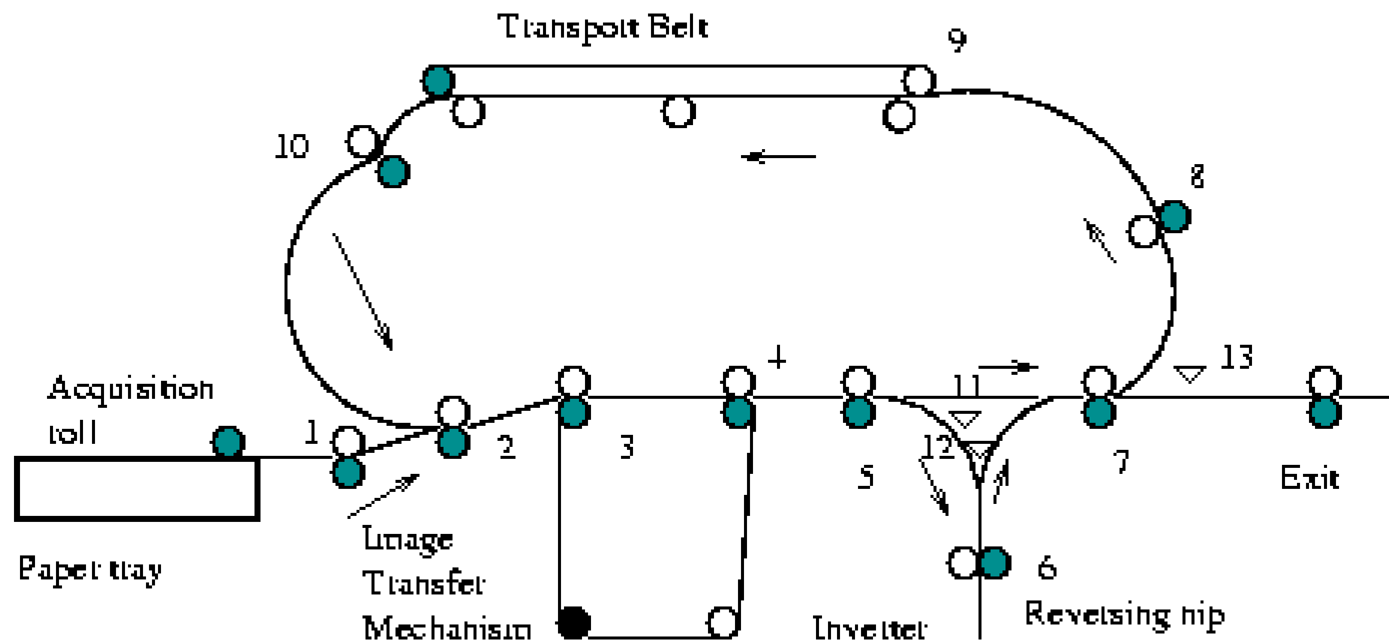


(From a paper by Fromherz et al. 2003)

Main goal: produce the documents as requested by the user

Rather than writing the control software, write a controller that produces and executes plans

The paper path



Monkeys can plan

Why shouldn't computers?

The problem statement: A monkey is in a laboratory room containing a box, a knife and a bunch of bananas. The bananas are hanging from the ceiling out of the reach of the monkey. How can the monkey obtain the bananas?

VHPOP coding of the monkey and bananas problem

```
(define (domain monkey-domain)
  (:requirements :equality)
  (:constants monkey box knife glass water
    waterfountain)
  (:predicates (on-floor) (at ?x ?y) (onbox ?x)
    (hasknife) (hasbananas) (hasglass) (haswater)
    (location ?x)

  (:action go-to
    :parameters (?x ?y)
    :precondition (and (not = ?y ?x)) (on-floor)
      (at monkey ?y)
    :effect (and (at monkey ?x) (not (at monkey ?y))))
```

VHPOP coding (cont'd)

```
(:action climb
  :parameters (?x)
  :precondition (and (at box ?x) (at monkey ?x))
  :effect (and (onbox ?x) (not (on-floor))))

(:action push-box
  :parameters (?x ?y)
  :precondition (and (not (= ?y ?x)) (at box ?y)
    (at monkey ?y) (on-floor))
  :effect (and (at monkey ?x) (not (at monkey ?y))
    (at box ?x) (not (at box ?y))))
```

VHPOP coding (cont'd)

```
(:action get-knife
  :parameters (?y)
  :precondition (and (at knife ?y) (at monkey ?y))
  :effect (and (hasknife) (not (at knife ?y))))

(:action grab-bananas
  :parameters (?y)
  :precondition (and (hasknife) (at bananas ?y)
    (onbox ?y) )
  :effect (hasbananas))
```

VHPOP coding (cont'd)

```
(:action pick-glass
  :parameters (?y)
  :precondition (and (at glass ?y) (at monkey ?y))
  :effect (and (hasglass) (not (at glass ?y))))

(:action get-water
  :parameters (?y)
  :precondition (and (hasglass) (at waterfountain ?y)
    (ay monkey ?y) (onbox ?y))
  :effect (haswater))
```

Problem 1: monkey-test1.pddl

```
(define (problem monkey-test1)
  (:domain monkey-domain)
  (:objects p1 p2 p3 p4)
  (:init (location p1) (location p2)
         (location p3) (location p4)
         (at monkey p1) (on-floor)
         (at box p2) (at bananas p3)
         (at knife p4))
  (:goal (hasbananas)))
```

go-to p4 p1
get-knife p4
go-to p2 p4
push-box p3 p2
climb p3
grab-bananas p3

time = 30 msec.

Problem 2: monkey-test2.pddl

```
(define (problem monkey-test2)
  (:domain monkey-domain)
  (:objects p1 p2 p3 p4 p6)
  (:init (location p1) (location p2)
         (location p3) (location p4) (location p6)
         (at monkey p1) (on-floor)
         (at box p2) (at bananas p3) (at knife p4)
         (at waterfountain p3) (at glass p6))
  (:goal (and (hasbananas) (haswater))))
```

go-to p4 p1
get-knife p4
go-to p6 p4
pick-glass p6
go-to p2 p6
push-box p3 p2
climb p3
get-water p3
grab-bananas p3

time = 70 msec.

Planning representation

Suppose that the monkey wants to fool the scientists, who are off to tea, by grabbing the bananas, but leaving the box in its original place. Can this goal be solved by a STRIPS-style system?

Comments on planning

- It is a synthesis task.
- Classical planning is based on the assumptions of a deterministic and static environment.
- Theorem proving and situation calculus are not widely used nowadays for planning (see below).
- Algorithms to solve planning problems include:
  - forward chaining: heuristic search in state space
  - Graphplan: mutual exclusion reasoning using plan graphs
  - Partial order planning (POP): goal directed search in plan space
  - Satifiability based planning: convert problem into logic

Comments on planning (cont'd)

- **Non-classical planners include:**
  - **probabilistic planners**
  - **contingency planners (a.k.a. conditional planners)**
  - **decision-theoretic planners**
  - **temporal planners**
  - **resource based planners**

Comments on planning (cont'd)

- **In addition to plan generation algorithms we also need algorithms for**
  - **Carrying out the plan**
  - **Monitoring the execution**
    (because the plan might not work as expected; or the world might change)
    (need to maintain the consistency between the world and the program's internal model of the world)
  - **Recovering from plan failures**
  - **Acting on new opportunities that arise during execution**
  - **Learning from experience**
    (save and generalize good plans)