

Classical Planning

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15-780 Graduate AI – Spring 2013

Readings:

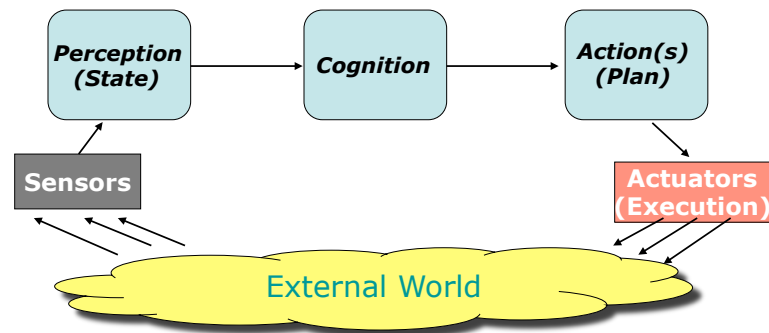
- Chapter 10, Russell & Norvig
- [Integrating planning and learning: The Prodigy architecture](#), (Sections 1 and 2)
M M. Veloso, J. Carbonell, M. Perez, D. Borrajo, E. Fink, and J. Blythe.
Journal of Experimental and Theoretical Artificial Intelligence, 7(1):81–120,
1995 (see pdf file off course website or www.cs.cmu.edu/~mmv/)

Planning – Problem Solving

Newell and Simon 1956

- Given the *actions* available in a task domain.
- Given a problem specified as:
 - an initial *state* of the world,
 - goal statement - a set of *goals* to be achieved.
- Find a *solution* to the problem, i.e., a way to transform the initial state into a new state of the world where the goal statement is true.
- Planning is “*thinking...*”

Intelligent Agents: Planning, Execution, and Learning



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Outline (2 lectures)

- Introduction: Search and Planning
- State, Actions, and Goal Representation
- Planning Algorithms
 - State-space Planning – GPS, Prodigy
 - Plan-space Planning – SNLP
 - GraphPlan
 - SATPlan
- Heuristics for Planning Algorithms
- Planning and Execution
 - Conditional Planning
 - Representation and algorithms
 - Information gathering actions
 - Replanning

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Planning - Search

- Search agent - issues
 - Atomic state representations
 - Instantiated actions
 - Domain-specific heuristics
 - Number of actions and states
- Planning agent
 - Factored state representation
 - Collection of variables
 - Actions schemas
 - Changes to the state

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Models of World State

- “Information-less” numerical identification (s1, s2,...)
- Symbolic – factored
 - Features
 - Predicates
- Conjunctive, enumerative, observable
- Complete, correct, deterministic
- Probabilistic, approximate, incremental, on-demand

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Classical Deterministic Planning

- Action Model
 - complete, deterministic, correct, STRIPS/PDDL representation, typed variables, CWA
- Single initial state, fully known
- Goal statement – set of goals

Several different planning algorithms

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STRIPS Representation

- Implicit Solution to Frame Problem
 - **State** is database of ground literals
 - If literal is not in database, assumed to be **false**
 - Effects of actions represented using **add** and **delete** lists (insert and remove literals from database)
 - No explicit representation of time
 - No logical inference rules
- Action representation
 - Conjunctive preconditions
 - Effects as **add** and **delete** lists

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The Blocks World - States

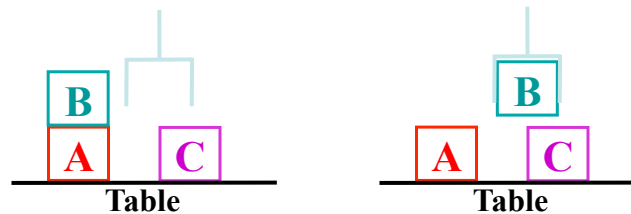
- Objects
 - Blocks: *A, B, C*
 - Table: *Table*
- Predicates
 - *On(A, B), On(C, Table), Clear(B), Handempty, Holding(C)*
 - *On-table(A), On(A,B), Top(B),...*
 - *Tower(A,B,C,...)*
- States – Conjunctive
 - *On(A,B) and On(B,C) and Clear(A) and Handempty*

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The Blocks World Definition – Actions



- Blocks are picked up and put down by the arm
- Blocks can be picked up only if they are clear, i.e., without any block on top
- The arm can pick up a block only if the arm is empty, i.e., if it is not holding another block, i.e., the arm can be pick up only one block at a time
- The arm can put down blocks on blocks or on the table

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Action Schema

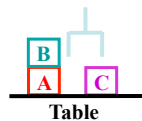
- Action Name
- All variables used in the schema
 - Universally quantified
 - Can choose values *to instantiate* the variables
- Precondition
- Effect
 - Positive effects – adds effect to state
 - Negative effects – deletes effects from state (if in state)

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STRIPS – The Blocks World

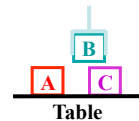


Pickup_from_table(b)
 Pre: Block(b), Handempty
 Clear(b), On(b, Table)
 Add: Holding(b)
 Delete: Handempty,
 On(b, Table)

Putdown_on_table(b)
 Pre: Block(b), Holding(b)
 Add: Handempty,
 On(b, Table)
 Delete: Holding(b)

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Pickup_from_block(b, c)
 Pre: Block(b), Handempty
 Clear(b), On(b, c), Block(c)
 Add: Holding(b), Clear(c)
 Delete: Handempty,
 On(b, c)

Putdown_on_block(b, c)
 Pre: Block(b), Holding(b)
 Block(c), Clear(c), $b \neq c$
 Add: Handempty, On(b, c)
 Delete: Holding(b), Clear(c)

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Actions

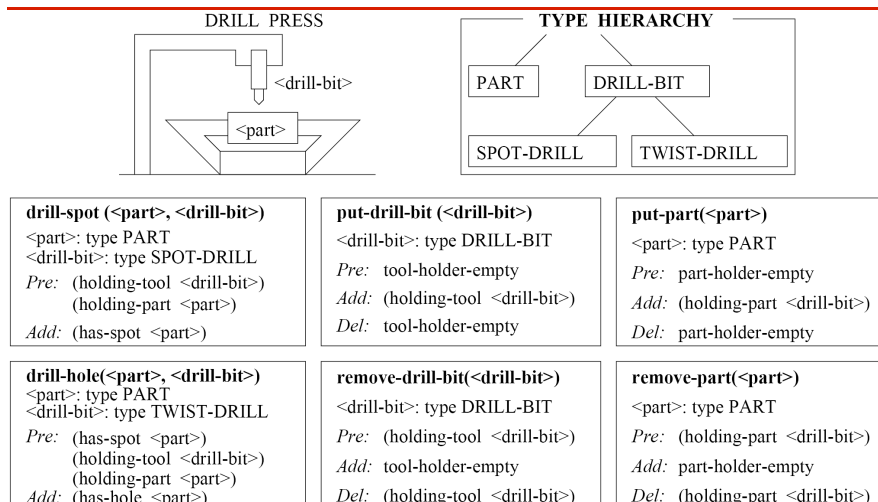
- An action a is **applicable** in s if all the preconditions of action a are satisfied by s .
- $\text{RESULT}(s,a) = (s - \text{Del}(a)) \cup \text{Add}(a)$
- No explicit mention of *time*
 - The precondition always refers to time t
 - The effect always refers to time $t+1$

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Example – Action Model

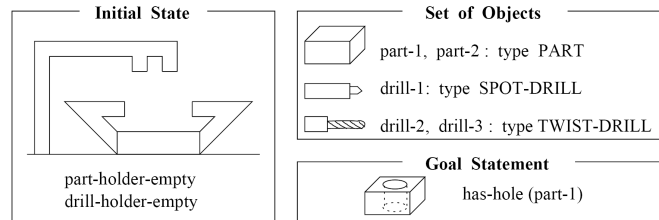


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Example – Problem and Plan



```

put-part(part-1)
put-drill=bit(drill-1)
drill-spot(part-1, drill-1)
remove-drill-bit(drill-1)
put-drill-bit(drill-2)
drill-hole(part-1, drill-2)

```

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Initial State, Goal, Actions Example-1

```

Init(At(C1, SFO) ∧ At(C2, JFK) ∧ At(P1, SFO) ∧ At(P2, JFK)
    ∧ Cargo(C1) ∧ Cargo(C2) ∧ Plane(P1) ∧ Plane(P2)
    ∧ Airport(JFK) ∧ Airport(SFO))
Goal(At(C1, JFK) ∧ At(C2, SFO))
Action(Load(c, p, a),
    PRECOND: At(c, a) ∧ At(p, a) ∧ Cargo(c) ∧ Plane(p) ∧ Airport(a)
    EFFECT: ¬ At(c, a) ∧ In(c, p))
Action(Unload(c, p, a),
    PRECOND: In(c, p) ∧ At(p, a) ∧ Cargo(c) ∧ Plane(p) ∧ Airport(a)
    EFFECT: At(c, a) ∧ ¬ In(c, p))
Action(Fly(p, from, to),
    PRECOND: At(p, from) ∧ Plane(p) ∧ Airport(from) ∧ Airport(to)
    EFFECT: ¬ At(p, from) ∧ At(p, to))

```

Figure 10.1 A PDDL description of an air cargo transportation planning problem.

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Questions

- Instantiated actions?
- Applicable actions?
- Result of applying an action?

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Domain and Actions

- A *domain* can be represented by many possible choices of literals, variables, actions, preconditions, effects.
- Choice of domain
 - Granularity of representation
 - Detail of reasoning
 - Effectiveness of search

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Initial State, Goal, Actions Example-2

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

Figure 10.3 A planning problem in the blocks world: building a three-block tower. One solution is the sequence [*MoveToTable*(C, A), *Move*($B, Table, C$), *Move*($A, Table, B$)].

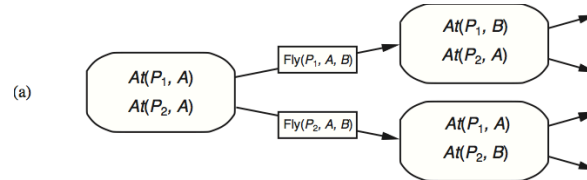
Domain Representation – Blocksworld

```

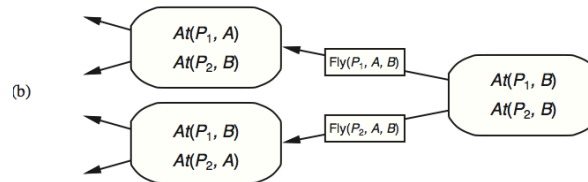
(OPERATOR MOVE
:preconds
  ?block BLOCK
  ?from OBJECT
  ?to OBJECT
  (and (clear ?block)
        (clear ?to)
        (on ?block ?from))
:effects
  add (on ?block ?to)
  del (on ?block ?from)
  (if (block-p ?from)
      add (clear ?from))
  (if (block-p ?to)
      del (clear ?to)))
  
```

Planning Algorithms

- Progression: Forward state-space search



- Regression: Backward state-space search

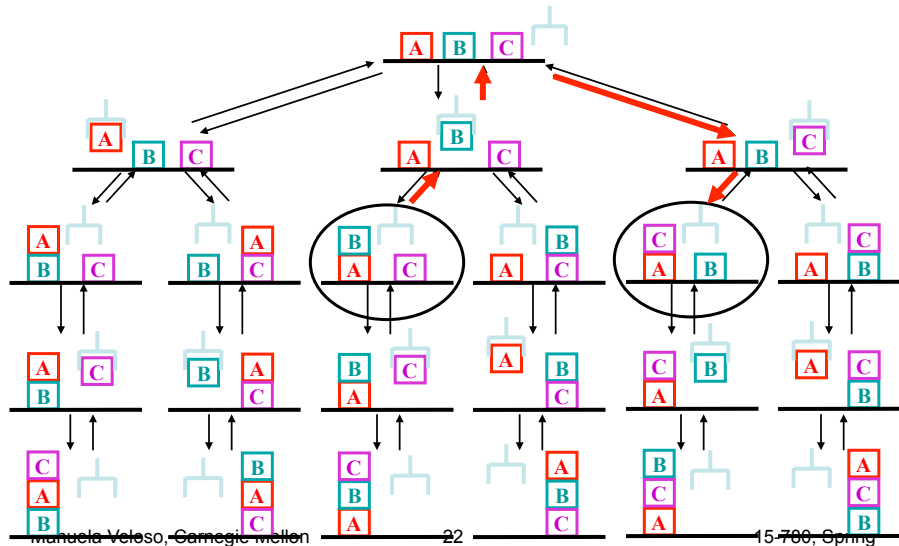


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Search Transitions Complete State



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Finding a Plan – Plan Generation

- **Backtracking Search** Through a *Search Space*
 - How to conduct the search
 - How to represent the search space
 - How to evaluate the solutions
- Non-Deterministic **Choice Points** Determine Backtracking
 - Choice of actions
 - Choice of variable bindings
 - Choice of temporal orderings
 - Choice of subgoals to work on

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Properties of Planning Algorithms

- **Soundness**
 - A planning algorithm is **sound** if all solutions are *legal* plans
 - All preconditions, goals, and any additional constraints are satisfied
- **Completeness**
 - A planning algorithm is **complete** if a solution can be found whenever one actually exists
 - A planning algorithm is **strictly complete** if all solutions are included in the search space
- **Optimality**
 - A planning algorithm is **optimal** if it maximizes a predefined measure of plan quality

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Linear Planning

- Basic Idea – Goal **stack**
 - *Work **on one goal until completely solved** before moving on to the next goal*

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Means-Ends Analysis

- Basic Idea
 - *Search by reducing the difference between the state and the goals*
 - What **means** (operators) are available to achieve the desired **ends** (goal)

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Means-ends Analysis in Linear Planning

(Newell and Simon 60s)

GPS Algorithm (*state*, *goals*, *plan*)

- If $goals \subseteq state$, then return (*state*, *plan*)
- **Choose** a difference $d \in goals$ between *state* and *goals*
- **Choose** an operator *o* to reduce the difference *d*
- If no applicable operators, then return *False*
- (*state*, *plan*) = **GPS** (*state*, preconditions(*o*), *plan*)
- If *state*, then return **GPS** (apply (*o*, *state*), *goals*, [*plan*, *o*])

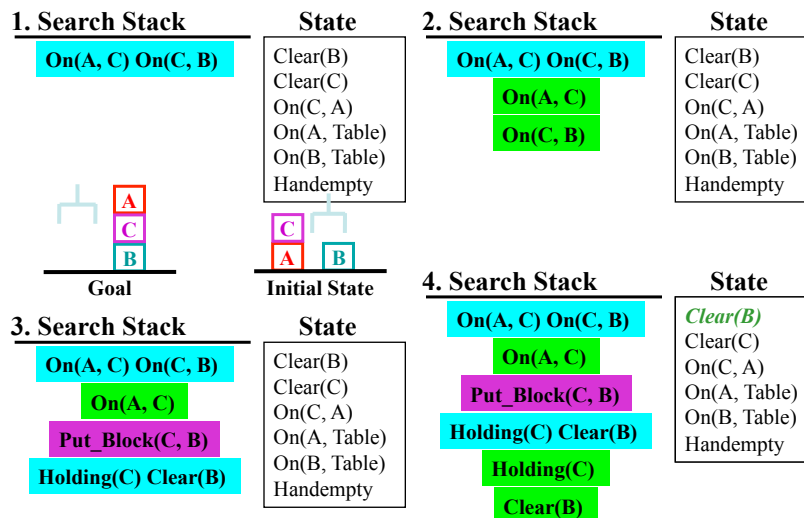
Initial call: GPS (initial-state, initial-goals, [])

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GPS Blocks-World Example



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GPS Blocks-World Example

5. Search Stack

On(A, C) On(C, B)

On(A, C)

Put_Block(C, B)

Holding(C) Clear(B)

Holding(C)

State

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

6. Search Stack

On(A, C) On(C, B)

On(A, C)

Put_Block(C, B)

Holding(C) Clear(B)

Pick_Block(C)

Handempty Clear(C) On(C, ?b)

State

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

7. Search Stack

On(A, C) On(C, B)

On(A, C)

Put_Block(C, B)

Holding(C) Clear(B)

Pick_Block(C)

State

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

8. Search Stack

On(A, C) On(C, B)

On(A, C)

Put_Block(C, B)

Holding(C) Clear(B)

[Pick_Block(C)]

State

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

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GPS Blocks-World Example

9. Search Stack

On(A, C) On(C, B)

On(A, C)

Put_Block(C, B)

State

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

[Pick_Block(C)]

10. Search Stack

On(A, C) On(C, B)

On(A, C)

State

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

[Pick_Block(C); Put_Block(C, B)]

11. Search Stack

On(A, C) On(C, B)

Put_Block(A, C)

Holding(A) Clear(C)

State

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

[Pick_Block(C)

Put_Block(C, B)]

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12. Search Stack

On(A, C) On(C, B)

Put_Block(A, C)

Holding(A) Clear(C)

Holding(A)

Clear(C)

State

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

[Pick_Block(C)

Put_Block(C, B)]

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GPS Blocks-World Example

13. Search Stack

On(A, C) On(C, B)
Put_Block(A, C)
Holding(A) Clear(C)
Holding(A)

State

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

[Pick_Block(C);
Put_Block(C, B)]

15. Search Stack

On(A, C) On(C, B)
Put_Block(A, C)
Holding(A) Clear(C)
Pick_Table(A)

State

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

[Pick_Block(C);
Put_Block(C, B)]

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14. Search Stack

On(A, C) On(C, B)
Put_Block(A, C)
Holding(A) Clear(C)
Pick_Table(A)
Handempty Clear(A)
On(A, Table)

State

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

[Pick_Block(C); Put_Block(C, B)]

16. Search Stack

On(A, C) On(C, B)
Put_Block(A, C)
Holding(A) Clear(C)

State

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

[Pick_Block(C);
Put_Block(C, B);
Pick_Table(A)]

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31

GPS Blocks-World Example

17. Search Stack

On(A, C) On(C, B)
Put_Block(A, C)

State

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

[Pick_Block(C);
Put_Block(C, B);
Pick_Table(A)]

18. Search Stack

On(A, C) On(C, B)

State

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

[Pick_Block(C);
Put_Block(C, B);
Pick_Table(A);
Put_Block(A, C)]

19. Search Stack

State

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

[Pick_Block(C);
Put_Block(C, B);
Pick_Table(A);
Put_Block(A, C)]

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32

Linear Planning with MEA

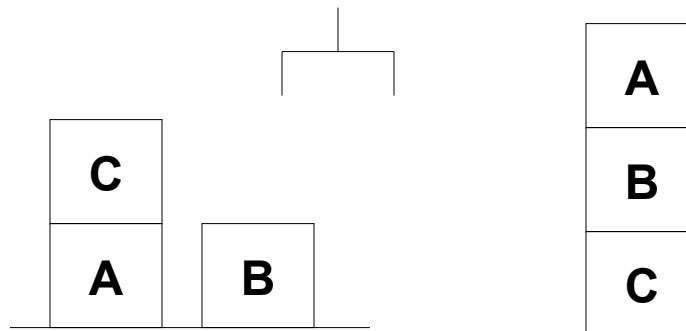
- Sound?
- Optimal?
- Complete?

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The Sussman Anomaly



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4-Action Blocks World Domain

Pickup (?b)

Pre: (handempty)
 (clear ?b)
 (on-table ?b)
 Add: (holding ?b)
 Delete: (handempty)
 (on-table ?b)
 (clear ?b)

Putdown (?b)

Pre: (holding ?b)
 Add: (handempty)
 (on-table ?b)
 Delete: (holding ?b)

Unstack (?a, ?b)

Pre: (handempty)
 (clear ?a) (on ?a ?b)
 Add: (holding ?a) (clear ?b)
 Delete: (handempty)
 (on ?a ?b) (clear ?a)

Stack (?a, ?b)

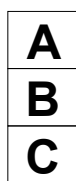
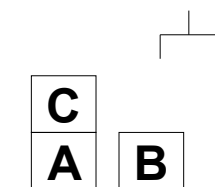
Pre: (holding ?a) (clear ?b)
 Add: (handempty)
 (on ?a ?b)
 Delete: (holding ?a)
 (clear ?b)

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The Sussman Anomaly



Linear Solution:

- (on B C)
 - Pickup (B)
 - Stack (B, C)
- (on A B)
 - Unstack (B, C)
 - Putdown (B)
 - Unstack (C, A)
 - Putdown (C)
 - Stack (A, B)
- (on B C)
 - Unstack (A, B)
 - Putdown (A)
 - Pickup (B)
 - Stack (B, C)
- (on A B)
 - Pickup (A)
 - Stack (A, B)

Linear Solution:

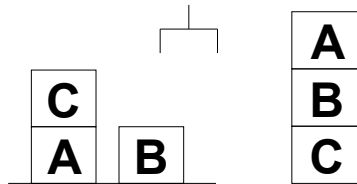
- (on A B)
 - Unstack (C, A)
 - Putdown (C)
 - Stack (A, B)
- (on B C)
 - Unstack (A, B)
 - Putdown (A)
 - Pickup (B)
 - Stack (B, C)
- (on A B)
 - Pickup (A)
 - Stack (A, B)

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NonLinear Solution – Optimal



NonLinear Solution:

- (on A B)
 - Unstack (C, A)
 - Putdown (C)
- (on B C)
 - Pickup (B)
 - Stack (B, C)
- (on A B)
 - Pickup (A)
 - Stack (A, B)

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Linear Planning – Goal Stack

- **Advantages**
 - Reduced search space, since goals are solved one at a time, and not all possible goal orderings are considered
 - Advantageous if goals are (mainly) independent
 - Linear planning is **sound**
- **Disadvantages**
 - Linear planning may produce **suboptimal** solutions (based on the number of operators in the plan)
 - Planner's efficiency is sensitive to goal orderings
 - Control knowledge for the “right” ordering
 - Random restarts
 - Iterative deepening
- Completeness?

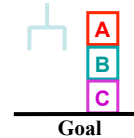
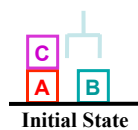
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Plan-Space, Partial-Order Planning

- General Approach
 - Find unachieved precondition
 - Add new action **or** connect to existing action
 - Determine if conflicts occur
 - Previously achieved precondition is “clobbered”
 - Fix conflicts (reorder, bind, ...)
- Partial-order planning can easily (and optimally) solve blocks world problems that involve goal interactions (e.g., the “Sussman Anomaly” problem)

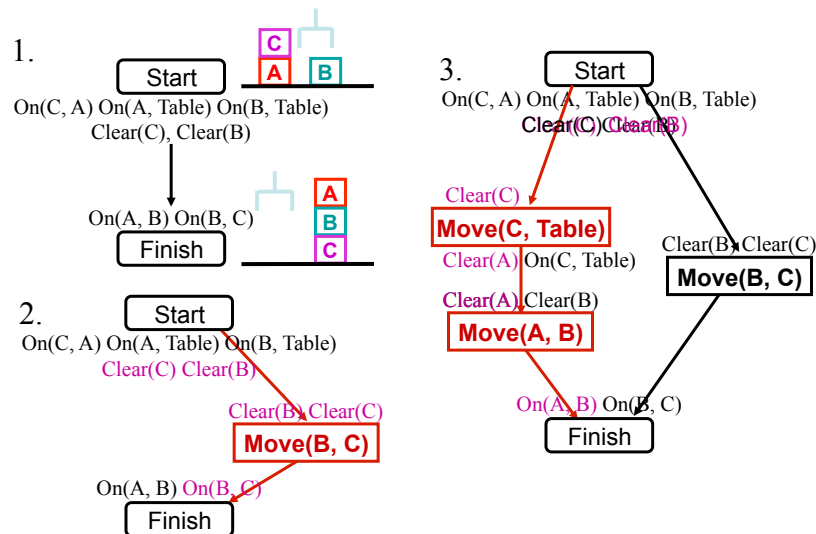


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POP and Sussman's Anomaly

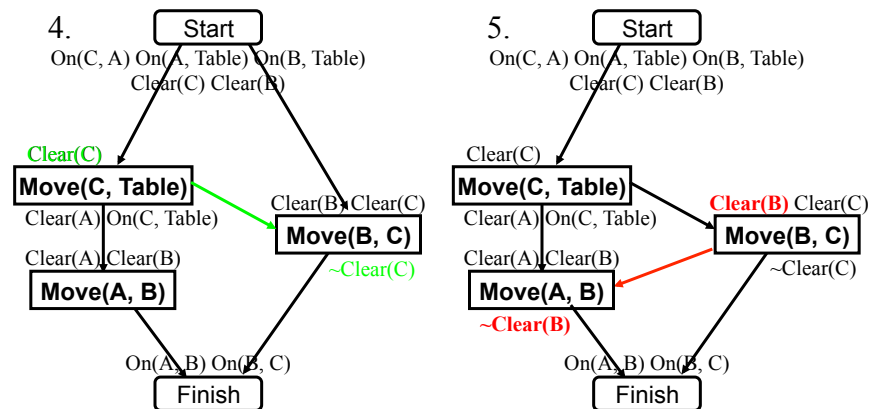


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POP and Sussman's Anomaly



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Least Commitment

- Basic Idea
 - *Make choices that are relevant only to solving the current part of the problem*
- Least Commitment Choices
 - **Orderings**: Leave actions unordered, unless they must be sequential
 - **Bindings**: Leave variables unbound, unless needed to unify with conditions being achieved
 - **Actions**: Usually not subject to “least commitment”

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POP – The Details

- **Causal Links**
 - Adding new actions or connecting to existing actions
 - The “purpose” of an action (which condition it supports)
 - $a_i \rightarrow^c a_j$, where a_i, a_j are actions and c is an effect of a_i
 - Plan = $\langle A, O, B, L \rangle$
- **Threats**
 - Finding and fixing conflicts
 - Action a_k with an effect c' that might “clobber” a causal link
 - **Promotion**: Order a_k after a_j
 - **Demotion**: Order a_k before a_i
 - **Separation**: Constrain c' so that it does not unify with c (non-codesignation constraint)

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Partial Order Planning: Discussion

- **Advantages**
 - Partial order planning is **sound** and **complete**
 - Typically produces better solutions (shorter), but not guaranteed **optimal**
 - Least commitment **may** lead to shorter search times
- **Disadvantages**
 - Significantly more complex algorithms (higher *per-node* cost)
 - Hard to determine what is true in a state
 - Larger search space (**infinite!**)

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44

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Plan Terminology

- **Totally Ordered** Plan
 - There exists sufficient orderings O such that all actions in A are ordered with respect to each other
- **Fully Instantiated** Plan
 - There exists sufficient constraints in B such that all variables are constrained to be equal to some constant
- **Consistent** Plan
 - There are no contradictions in O or B
- **Complete** Plan
 - Every precondition p of every action a_i in A is *achieved*:
There exists an effect of an action a_j that comes before a_i and unifies with p , and no action a_k that deletes p comes between a_j and a_i
- Partial plans are typically *not* executable *nor* consistent

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45

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Example: One-Way Rocket (Veloso 89)

<pre>(OPERATOR LOAD-ROCKET :preconds ?roc ROCKET ?obj OBJECT ?loc LOCATION (and (at ?obj ?loc) (at ?roc ?loc)) :effects add (inside ?obj ?roc) del (at ?obj ?loc))</pre>	<pre>(OPERATOR UNLOAD-ROCKET :preconds ?roc ROCKET ?obj OBJECT ?loc LOCATION (and (inside ?obj ?roc) (at ?roc ?loc)) :effects add (at ?obj ?loc) del (inside ?obj ?roc))</pre>	<pre>(OPERATOR MOVE-ROCKET :preconds ?roc ROCKET ?from-l LOCATION ?to-l LOCATION (and (at ?roc ?from-l) (has-fuel ?roc)) :effects add (at ?roc ?to-l) del (at ?roc ?from-l) del (has-fuel ?roc))</pre>
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Incompleteness of Linear Planning

Initial state: Goal statement:

```

(at obj1 locA)
(at obj2 locA)
(at ROCKET locA)
(has-fuel ROCKET)

```

```

(and
  (at obj1 locB)
  (at obj2 locB))

```

<i>Goal</i>	<i>Plan</i>
(at obj1 locB)	(LOAD-ROCKET obj1 locA) (MOVE-ROCKET) (UNLOAD-ROCKET obj1 locB)
(at obj2 locB)	<i>failure</i>

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47

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State-Space Nonlinear Planning

Extend linear planning:

- From **stack** to **set** of goals.
- Include in the search space all possible interleaving of goals.

State-space nonlinear planning is **complete**.

<i>Goal</i>	<i>Plan</i>
(at obj1 locB)	(LOAD-ROCKET obj1 locA)
(at obj2 locB)	(LOAD-ROCKET obj2 locA)
(at obj1 locB)	(MOVE-ROCKET) (UNLOAD-ROCKET obj1 locB)
(at obj2 locB)	(UNLOAD-ROCKET obj1 locB)

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48

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Prodigy Planner

- Extension to GPS
 - Set of goals, instead of stack of goals
 - Means-ends analysis for selection of “pending goals”
 - Choice point for applying an operator when applicable and continue backward-chaining (subgoalting)

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Prodigy4.0 (Veloso et al. 90)

1. Terminate if the goal statement is satisfied in the current state.
2. Compute the **SET** of *pending goals* G , and the **set** of *applicable operators* A .
 - A goal is pending if it is a precondition, not satisfied in the current state, of an operator already in the plan.
 - An operator is applicable when all its preconditions are satisfied in the state.
1. Choose a goal G in G or choose an operator A in A .

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50

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Prodigy4.0 Planning Algorithm

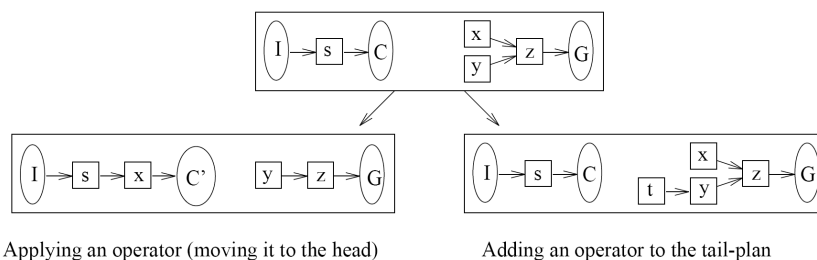
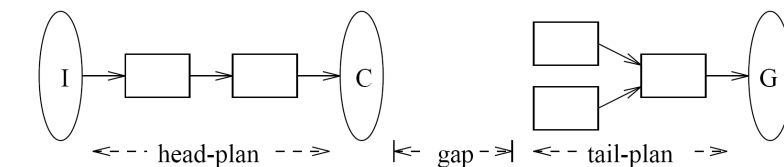
4. If G has been chosen, then
 - *Expand goal G ,*
i.e., get the set O of *relevant instantiated operators* that could achieve G ,
 - Choose an operator O from O ,
 - Go to step 1.
5. If an operator A has been selected as directly applicable, then
 - *Apply A ,*
 - Go to step 1.

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51

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Prodigy4.0 – Search Representation



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Why is Planning Hard?

Planning involves a complex search:

- Alternative operators to achieve a goal
- Multiple goals that interact
- Solution optimality, quality
- Planning efficiency, soundness, completeness

Many Issues in Planning

- State representation
 - The frame problem
 - The “choice” of predicates
 - On-table (x), On (x, table), On-table-A, On-table-B,...
- Action representation
 - Many alternative definitions
 - Reduce to “needed” definition
 - Conditional effects
 - Uncertainty
 - Quantification
 - Functions
- Generation – planning algorithm(S)

Many Planning “Domains”

- Web management agents
- Robot planning
- Manufacturing planning
- Image processing management
- Logistics transportation
- Crisis management
- Bank risk management
- Blocksworld
- Puzzles
- Artificial domains

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55

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Summary

- **Planning:** selecting one sequence of actions (operators) that transform (apply to) an initial state to a final state where the goal statement is true.
- **Means-ends analysis:** identify and reduce, as soon as possible, *differences* between state and goals.
- **Linear planning:** backward chaining with means-ends analysis using a stack of goals - potentially efficient, possibly unoptimal, incomplete; GPS, STRIPS.
- **Plan-space planning:** least commitment search; causal links and threat resolution; effective management of goal interactions
- **Nonlinear planning with means-ends analysis:** backward chaining using a set of goals; reason about *when* “to reduce the differences;” Prodigy4.0.

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56

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