

Plan Generation Classical Planning

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Outline

- What is a *State and Goal*
- What is an *Action*
- **What is a *Plan***
- *Finding* a Plan

What is a Plan?

- *Sequence* of Instantiated Actions
- *Partial Order* of Instantiated Actions
- *Set* of Instantiated Actions
- *Policy*
 - Mapping from states to actions

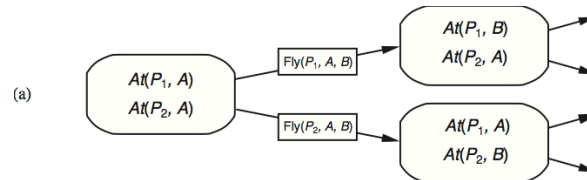
Increasing Generality
↓

Outline

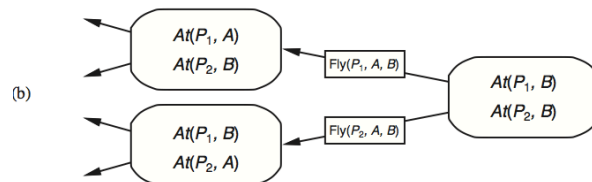
- What is a *State and Goal*
- What is an *Action*
- What is a *Plan*
- ***Finding a Plan***

Planning Algorithms

- Progression: Forward state-space search



- Regression: Backward state-space search



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

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

Linear Planning

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

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)

Means-ends Analysis in Linear Planning

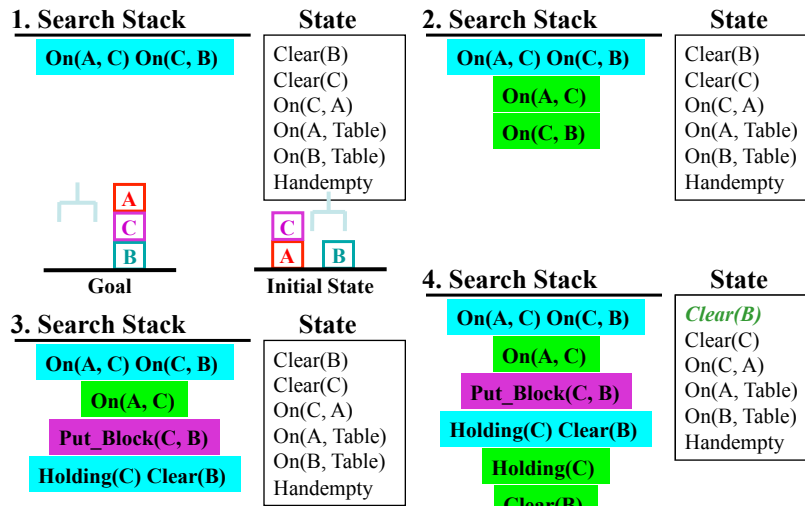
(Newell and Simon 60s)

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

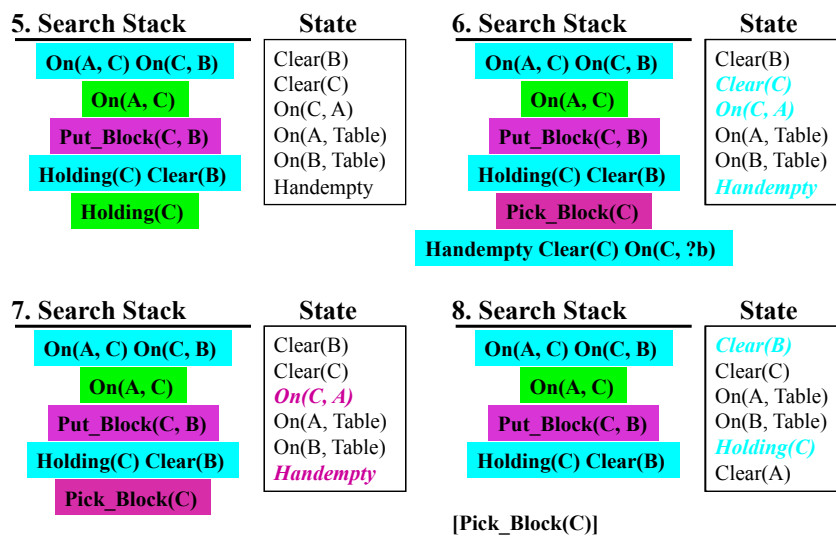
- If $goals \subseteq state$, then return (*state*, *plan*)
- **Choose** a difference $d \in goals$ between *state* and *goals*
- **Choose** an operator o to reduce the difference d
- If no applicable operators, then return *False*
- (*state*, *plan*) = **GPS** (*state*, $preconditions(o)$, *plan*)
-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

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

GPS Blocks-World Example



GPS Blocks-World Example



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)]

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)]

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)]

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)]

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)]

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)]

GPS Blocks-World Example

17. Search Stack

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

Put_Block(A, C)

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

State

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

18. Search Stack

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

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

State

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

19. Search Stack

State

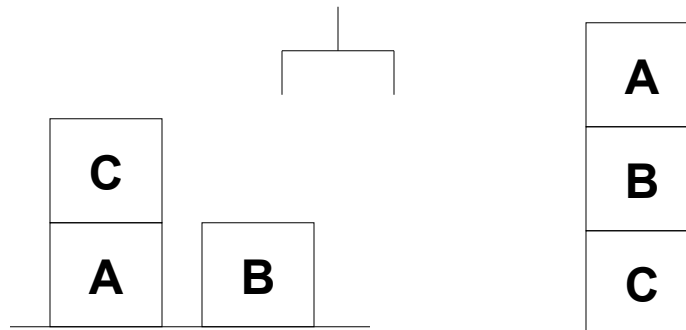
On(B, Table)
Clear(A)
On(C, B)
Handempty
On(A, C)

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

Linear Planning with MEA

- Sound?
- Optimal?
- Complete?

The Sussman Anomaly



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)

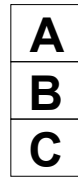
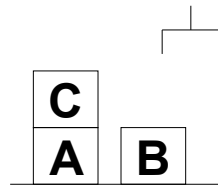
Unstack (?a, ?b)

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)

The Sussman Anomaly



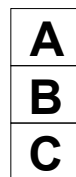
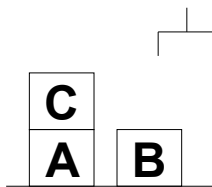
Linear Solution:

- (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)

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)

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?

Example: One-Way Rocket (Veloso 89)

```

(OPERATOR LOAD-ROCKET          (OPERATOR UNLOAD-ROCKET      (OPERATOR MOVE-ROCKET
:preconds                    :preconds                    :preconds
?roc ROCKET                  ?roc ROCKET                  ?roc ROCKET
?obj OBJECT                  ?obj OBJECT          ?from-1 LOCATION
?loc LOCATION                ?loc LOCATION        ?to-1 LOCATION
(and (at ?obj ?loc)          (and (inside ?obj ?roc) (and (at ?roc ?from-1)
(at ?roc ?loc))              (at ?roc ?loc))        (has-fuel ?roc))
:effects                      :effects
add (inside ?obj ?roc)        add (at ?obj ?loc)      add (at ?roc ?to-1)
del (at ?obj ?loc)           del (inside ?obj ?roc)  del (at ?roc ?from-1)
                                del (has-fuel ?roc)

```

Incompleteness of Linear Planning

Initial state: Goal statement:

```
(at obj1 locA)                      (and
(at obj2 locA)                      (at obj1 locB)
(at ROCKET locA)                   (at obj2 locB))
(has-fuel ROCKET)
```

<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>

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)

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 (subgoaling)

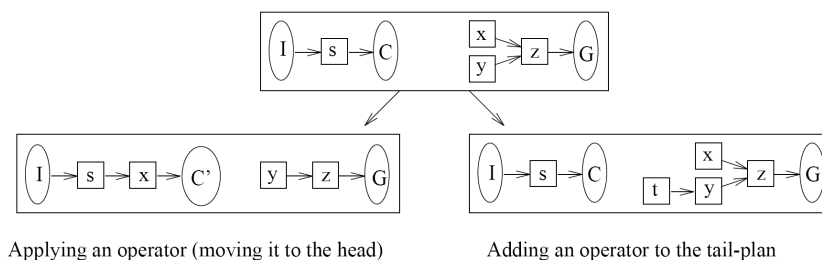
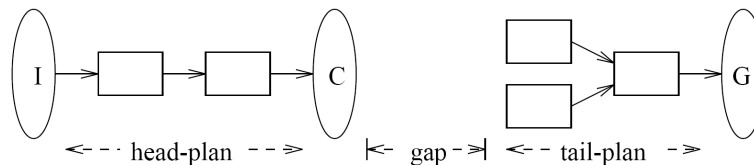
Prodigy4.0 (Veloso et al. 90)

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

Prodigy4.0 Planning Algorithm

4. If the pending goal 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 , as a *relevant operator* for goal G .
 - Go to step 1.
5. If a relevant operator A has been selected as directly applicable, then
 - *Apply A* ,
 - Go to step 1.

Prodigy4.0 – Search Representation



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

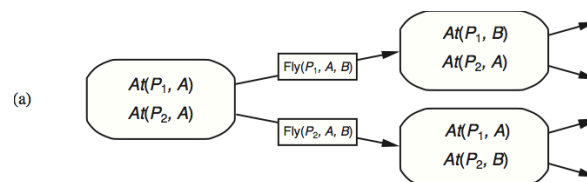
- State representation
 - The frame problem
 - The “choice” of predicates
 - On-table (x), On (x, table), On-table-A, On-table-B,...
- Action representation
 - Many alternative definitions
 - Reduce to “needed” definition
 - Conditional effects
 - Uncertainty
 - Quantification
 - Functions
- Generation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

Summary

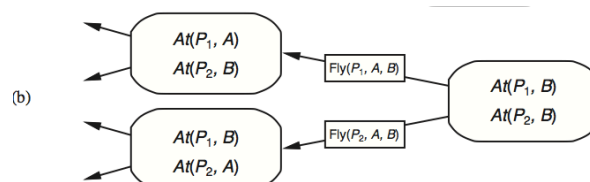
- **Planning:** selecting one sequence of actions (operators) that transform (apply to) an initial state to a final state where the goal statement is true.
- **Means-ends analysis:** identify and reduce, as soon as possible, *differences* between state and goals.
- **Linear planning:** backward chaining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

Planning Algorithms

- **Progression: Forward state-space search**

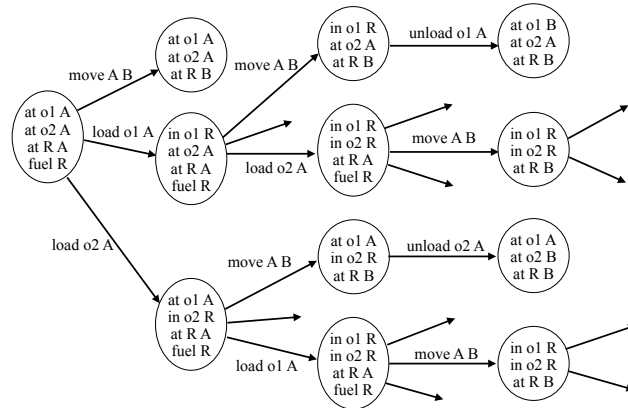


- **Regression: Backward state-space search**



Planning Graph – Forward Expansion

- State reachability – “until” goal
 - Can find **all** goals reachable from initial state
 - Exponential in time and memory



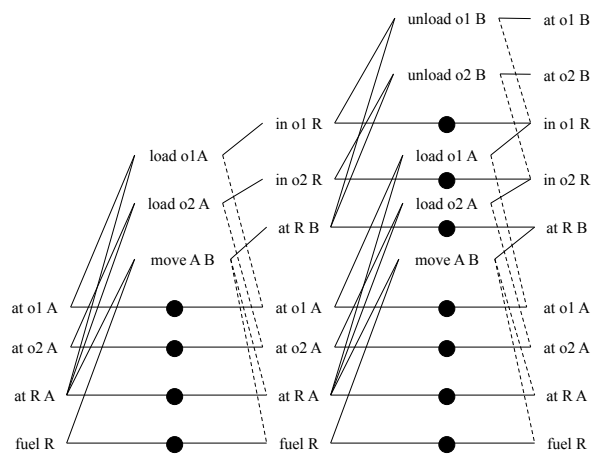
Graphplan

Blum & Furst 95

- Preprocessing before engaging in search.
- Forward search combined with backward search.
- Construct a *planning graph* to reveal constraints
- Two stages:
 - **Extend:** One time step in the planning graph.
 - **Search:** Find a valid plan in the planning graph.
- Graphplan finds a plan or proves that no plan has fewer “time steps.”

Plan Graph

One-Way Rocket Example



Extending a Planning Graph - Actions

- To create an action-level i :
 - Add each instantiated operator, for which all of its preconditions are present at proposition-level i AND *no two of its preconditions are exclusive*.
 - Add all the no-op actions.
- Determine the **exclusive** actions.

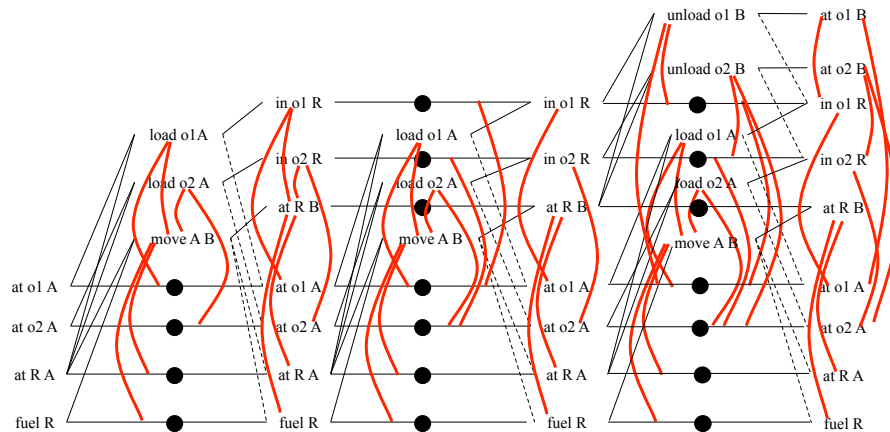
Extending a Planning Graph – Propositions

- To create a proposition-level $i + 1$:
 - Add all the effects of the inserted actions at action-level i - distinguishing add and delete effects.
- Determine the **exclusive** actions.

Planning Graphs

- A literal may exist at level $i + 1$ if it is an Add-Effect of some action in level i .
- Two propositions p and q are **exclusive** in a proposition-level if ALL actions that add p are exclusive of ALL actions that add q .
- Actions A and B are **exclusive** at action-level i , if:
 - **Interference**: A (or B) deletes a precondition or an Add-Effect of B (or A).
 - **Competing Needs**: p is a precondition of A and q is a precondition of B, and p and q are exclusive in proposition-level $i - 1$.

Mutex Exclusivity Relations One-Way Rocket Example



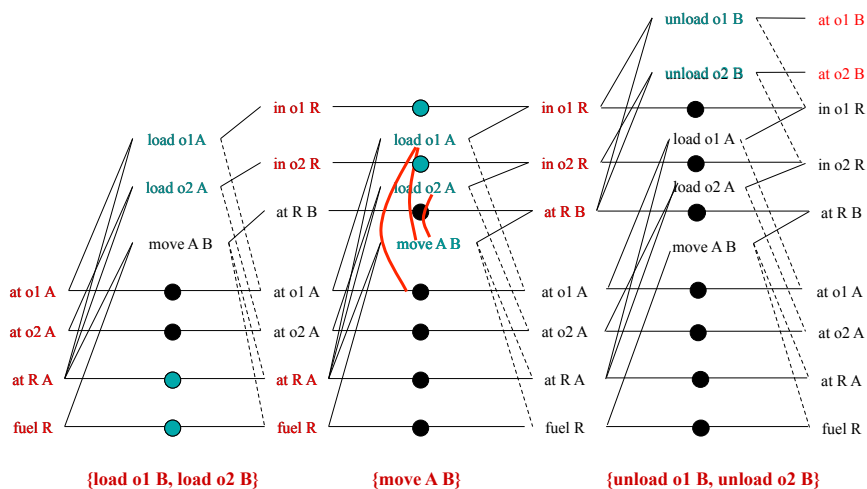
Exclusivity Examples

- Exclusive Actions: (Move A B) deletes a precondition of (Load o1 A). Therefore exclusive (existence of threats).
- Exclusive Propositions: (at R A) and (at R B) at time 2 are exclusive. (at R A) is added by a no-op and (at R B) is added by (Move A B) and no-op and (Move A B) are exclusive actions.
- Exclusive Actions: Then (Load o1 A) and (Load o2 B) are exclusive because (at R A) and (at R B) are exclusive.
- Propositions can be exclusive in some time step and not in others: If (at o1 A) and (at R A) at time 1, then (in o1 A) and (at R B) are exclusive at time 2, but not at time 3.

Searching a Planning Graph

- Level-by-level backward-chaining approach to use the exclusivity constraints.
- Given a set of goals at time t , identify all the sets of actions (including no-ops) at time $t - 1$ who add those goals and are not exclusive. The preconditions of these actions are new goals for $t - 1$.

Searching a Planning Graph



Recursive Search

- For each goal at time t in some arbitrary order:
 - Select some action at time $t - 1$ that achieves that goal and it is not exclusive with any other action already selected.
 - Do this recursively for all the goals at time t - do not add new action, but use the ones already selected if they add another goal.
 - If recursion returns failure, then select a different action.
- The new goal set is the set of all the preconditions of the selected actions.

Enhancements

- Forward-checking - for the goals ahead, check if all the actions that add it are exclusive with the selected action.
- Memoization - when a set of goals is not solvable at some time t , then this is recorded and hashed. If back at time t , the hash table is checked and search proceeds backing up right away.

Planning as Satisfiability

- One interpretation: ``first-order deductive theorem-proving does not scale well.'`
- One solution: ``propositional satisfiability'`
- Uniform clausal representation for goals and operators.
- Stochastic local search is a powerful technique for planning.