Planning

- Planning involve reasoning about actions that the agent intends to carry out.
 - Planning is the reasoning side of actions.
 - Reasoning involves the representation of the world that the agent has and the representation of its own actions.
 - Agent plans for actions and executes them to achieve the goal.
- The planning problem can be described at varying levels of detail.
 - Simplest planning to complex planning

Simplest planning problems

- In the simplest planning problems
 - The domain is static
 - The agent has complete information of domain (perception is perfect)
 - Agent can take all facts into account while planning
 - Actions are instantaneous
 - There is no notion of time, but only of sequencing of actions.
 - Actions effects are deterministic
 - Agent is the only one changing the world.

Complex planning

- Each of these constraints can be relaxed to define richer planning domains.
- In these problems planning is a computationally harder task.

- Planning requires explicit representation of actions in a domain.
- This is little different from our view of search problems that assume a move generation function.
- Representation actions explicitly allows us to explore richer domains for example when actions have durations.
- So it requires other approaches to planning different from graph search.

Classical planning

- Reasoning about actions involves simulation of changes in the mental model (representation of the world)
- Agent peers into the future to examine the different possibilities and select the desirable course of action.
 - This is called projection into future.
 - The planner searches through the possible combination of actions to find the plan that will work.

Alternative Approach: Memory based planning

- To exploit agent experience, looking into past.
- If the agent has memory, it can reuse
 - a plan that worked earlier to find a solution
 - Or problem solving experience to find a solution
- It involves representation and reasoning over the events in the past.
- The idea is
 - If a problem is familiar, then a known solution is used from memory
 - Is a problem is new, earlier project based methods have to be used.
- This is more intelligent agent: learning from the past.

Planning Domain Description Language (PDDL)

- A standard notation for describing the world and actions
- Allows to express world in terms of the language and actions in terms of schemas
- The language is based on First Order Logic Notation
- It also allows to state a specific planning problem by describing the initial state and the goals using the constructs define in the schema
- The planning algorithms can be written in a domain independent form.
- The domain and problem descriptions will become input to the domain independent planning algorithms.

- The complexity of planning domain will depend on the expressiveness of the PDDL and the constructs defined.
- Starting from simplest planning domain and problems, a series of more expressive PDDL languages have been defined.
- In each new languages, the world and actions can be described in richer fashion, taking more and more aspects from the real world

STRIPS domains

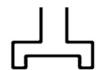
- Simplest of planning domains
 - Pose problems that are computationally hard to solve
- They are basically viewed as search problems in the classical approaches.

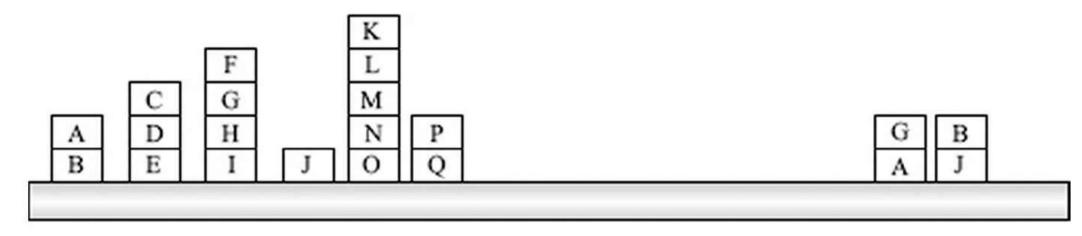
The STRIPS Domain

- Search algorithms can also be used for planning
- World is described in states and actions are described as operators
- Operators are applied to a state and transforms into another state
- The operators are close to the idea of production rules being applicable when the associated preconditions hold.
- The first program to use this formalism was STRIPS

The STRIPS Domain

- The program was designed to plan in the domain of block world.
- In this domain a set of labelled blocks are in some configuration to start with.
- A block can be on another block, or a block can be on the table and the table has unlimited capacity.
- The objective of the planner is to rearrange the blocks into a desired (goal) configuration
- The output of the planner is the set of moves that brings about this desired rearrangement





The given state

The goal = on(G, A) Λ on(B, J)

- The representation borrows from logic, specifically first order logic
- Unlike in classical logic, predicates in planning such as on(C,D) can be true at one time instance and false at another.
 - Such predicates are also known as fluents
- Classical reasoning with first order logic does not allow for assertions to be withdrawn.
- One way to handle the problem is to introduce time as a parameter and have every predicate timestamped.
- However this creates an enormous book-keeping problem of carrying forward facts as time moves on.

- How do we know that if a fact like ontable(B, t1) is true the ontable(B, t2) will be true as well?
- This is well known frame problem
- This problem can be solved by adding the so called "frame axioms" which explicitly specify that all conditions not affected by actions are not changed while executing that action.
- Thus the frame axioms need to be applied at each stage to keep track
 of what is true as time marches on

- The ingenious solution introduced in STRIPS was to make all changes explicit
 - The assertions that became true as a consequence of the operator being applied will be added,
 - The assertions that became false as a consequence of the operator being applied will be deleted,
 - Whatever was not changed remained the same.

- STRIP operators follow this strategy and are made of the following three components
 - P: Precondition List
 - The assertions needed to be true for the operator to be applicable
 - A: Add List
 - The assertions that became true as a consequence of the operator being applied will be added,
 - D: Delete List
 - The assertions that became false as a consequence of the operator being applied will be deleted,

The World in the block world domain

- On(X, Y): Block X is on block Y
- Ontable(X): block X is on the table
- Clear(X): no block is on block X
- Holding(X): the robot arm is holding X
- Armempty: the robot arm is not holding anything

Operators for the blocks world domain

• PICKUP(X)

- P: Ontable(X) ^ Clear(X) ^ Armempty
- A: Holding(X)
- D: Ontable(X) ^ Armempty

PUTDOWN(X)

- P: Holding(X)
- A: Ontable(X) ^ Armempty
- D: Holding(X)

Operators for the blocks world domain

- UNSTACK(X,Y)
 - P: ON(X, Y) ^ clear (X) ^ armempty
 - A: holding(X) ^ clear(Y)
 - D: on(X,Y) ^ armempty
- STACK(X,Y)
 - P: holding(X) ^ clear(Y)
 - A: ON(X, Y) ^ clear (X) ^ armempty
 - D: : holding(X) ^ clear(Y)

- Since goal state description may not describe the final state completely
 - There can be many states in which the given goal conditions will be true
- This also implies that backward reasoning will have to deal with incomplete state descriptions.

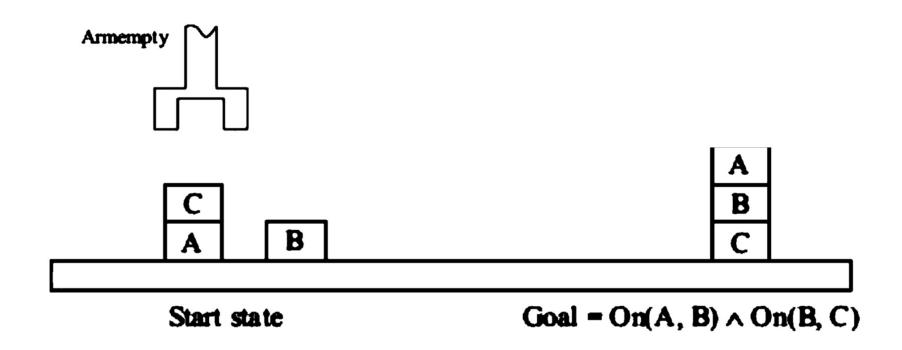
Planning Algorithms in the STRIP domain

- The search algorithms can easily be adapted to the task of planning
 - Known as forward reasoning/search
- We can start the search from the goal description
 - Known as backword reasoning/search
- We can combine the good features of both, forward and backward search
 - Known as goal stack planning
- We can seek to fill in the actions constituting the plan in a nonlinear order, filling in actions as and when we spot their requirements
 - Plan space planning and Partial order planning
- Finally, hierarchical planning

Forward state space planning

- Given a state, and given a set of operations, the actions can be determined that can be applied in the state to generate successor states.
- This corresponds to implement MovGen function

Planning problem from domain of blocks world



Same problem described in PDDL

- Lets effect(a) denotes all the effects of action a.
- Lets effect⁺(a) denotes the set of positive effects of action a
- This is the same as the add list in the STRIPS notation
- Corresponding to the delete list, we have effect⁻(a)
- Then given as state S in which the action a is applicable, the next state
 S' after the action a is applied is given by

$$S' \leftarrow \{S - effects^-(a)\} \cup effects^+(a)$$

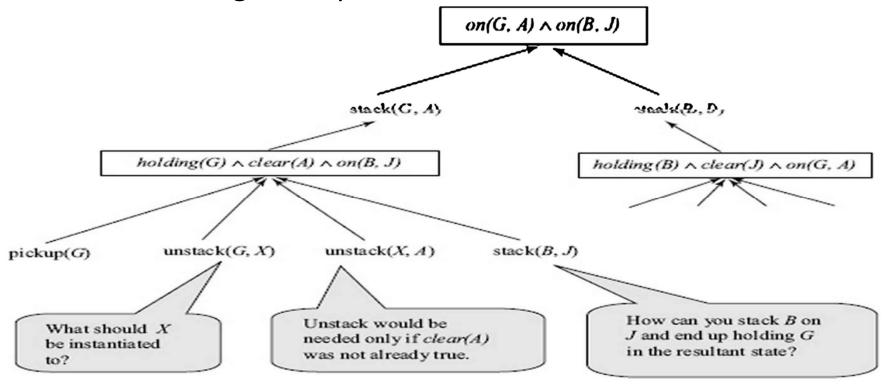
- We say that the state has progressed through the action a.
- Let us define a function Progress(A, S) that returns the successor state when action A is applied to S
- Given the progress(A,S) function, the search algorithms can be used.
- However they will suffer with same drawbacks
 - The search space generated will be huge
 - The search algorithm will have no sense of direction

Backward state space planning

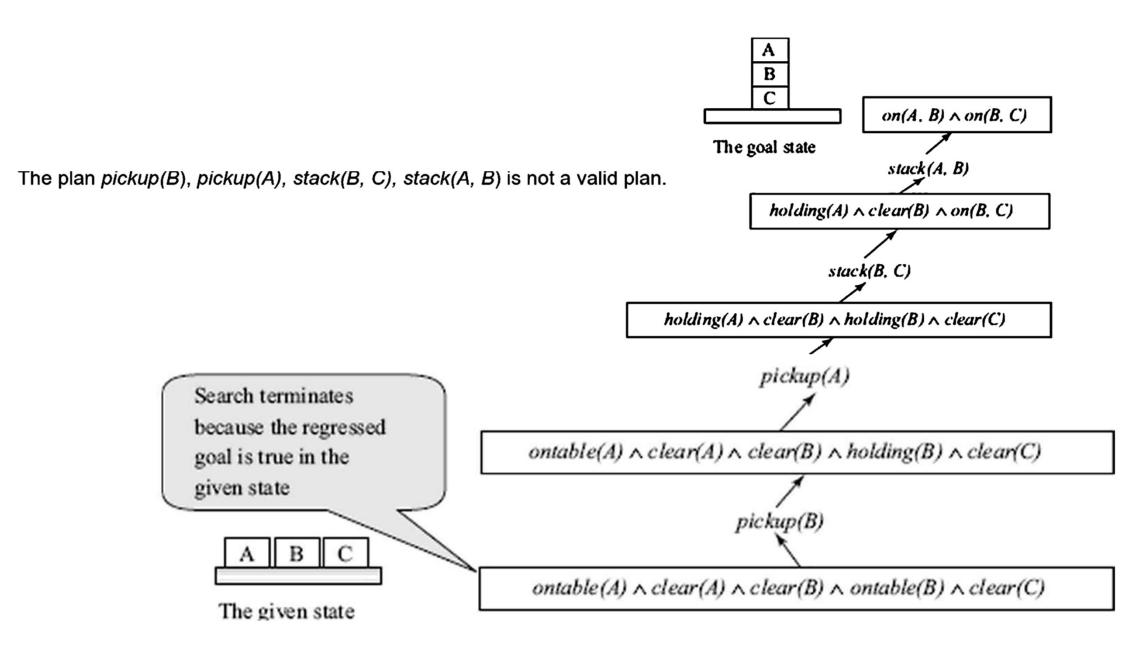
- FSSP is not knowing the goal and simply explores the set of all future states possible in some predetermined order.
- Backward Goal based reasoning is fundamental to intelligent behaviour
- It is also the foundation of reasoning with logic
- For backward planning we first need to define the regressionopposite of progression
 - It allows us to move back from set of goal clauses to a set of subgoal clauses

- The regressed goal represents the minimum set of facts that must be true in a state in which the action can be applied and which would result in a goal state.
- Form the set of goal facts, remove the effect of the action and add the preconditions of the action
- Lets define a function regress(A,G) that returns the regressed goal over action A when applied to goal G
- The regresses goal represents the minimum set of facts that must be true in a state in which action can be applied and which would result an goal state

However the regressor process is not sound



- The relevance of action to a goal simply means that it looks like that the action could achieve some part of the goal.
- But it does not mean that the action may be applicable in the preceding state.



One can generate plan and check whether it is a valid plan or not