

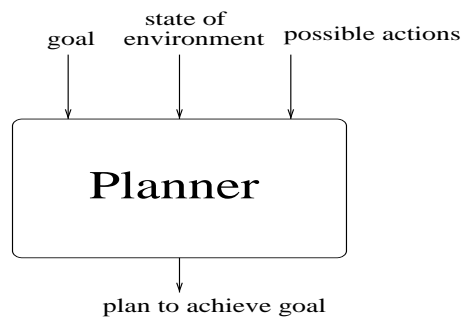
LINEAR PLANNING

1 What is Planning?

- Key problem facing *agent* is *deciding what to do*.
- We want agents to be *taskable*: give them *goals* to achieve, have them decide for themselves how to achieve them.
- Basic idea is to give an agent:
 - representation of goal to achieve;
 - knowledge about what actions it can perform; and
 - knowledge about state of the world;and to have it generate a *plan* to achieve the goal.

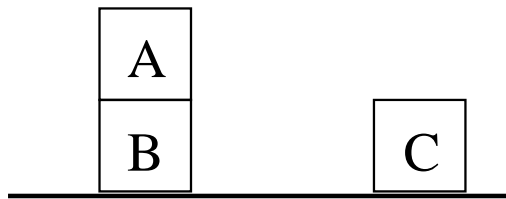
- Essentially, this is

automatic programming.



- Question: How do we *represent...*
 - goal to be achieved;
 - state of environment;
 - actions available to agent;
 - plan itself.
- We show how all this can be done in first-order logic.
- This isn't the only way to solve the problem, and later we'll look at other approaches.

- We'll illustrate the techniques with reference to the *blocks world*.
- Contains a robot arm, 3 blocks (A, B and C) of equal size, and a table-top.
- Initial state:



- Though this is a toy problem, it is a good place to start thinking about planning.

- To represent this environment, need an *ontology*.

$On(x, y)$ obj x on top of obj y
 $OnTable(x)$ obj x is on the table
 $Clear(x)$ nothing is on top of obj x
 $Holding(x)$ arm is holding x

- We will also have *armEmpty* which we will use as abbreviation for:

$\neg \exists x, Holding(x)$

meaning that there is no object x that is being held by the arm.

- Here is a first order logic (FOL) representation of the blocks world described above:

$Clear(A)$
 $On(A, B)$
 $OnTable(B)$
 $OnTable(C)$
 $Clear(C)$

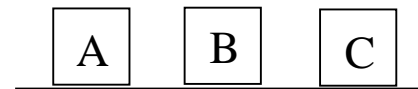
- Use the *closed world assumption*: anything not stated is assumed to be *false*.

- A *goal* is represented as a FOL formula.

- Here is a goal:

$OnTable(A) \wedge OnTable(B) \wedge OnTable(C)$

- Which corresponds to the state:



- *Actions* are represented using a technique that was developed in the STRIPS planner.

- Each action has:

- a *name*
which may have arguments;
- a *pre-condition list*
list of facts which must be true for action to be executed;
- a *delete list*
list of facts that are no longer true after action is performed;
- an *add list*
list of facts made true by executing the action.

Each of these may contain *variables*.

- Example 1:

The *stack* action occurs when the robot arm places the object *x* it is holding is placed on top of object *y*.

```
Stack(x, y)
pre Clear(y) ∧ Holding(x)
del Clear(y) ∧ Holding(x)
add ArmEmpty ∧ On(x, y)
```

- Example 2:

The *unstack* action occurs when the robot arm picks an object *x* up from on top of another object *y*.

```
UnStack(x, y)
pre On(x, y) ∧ Clear(x) ∧ ArmEmpty
del On(x, y) ∧ ArmEmpty
add Holding(x) ∧ Clear(y)
```

Stack and UnStack are *inverses* of one-another.

- Example 3:

The *pickup* action occurs when the arm picks up an object *x* from the table.

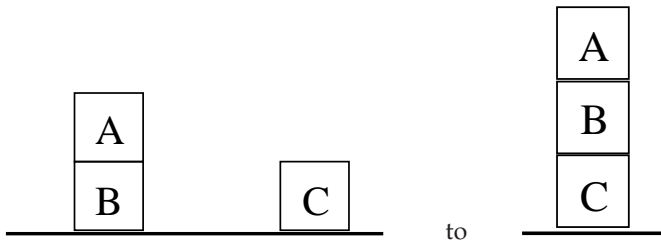
```
Pickup(x)
pre Clear(x) ∧ OnTable(x) ∧ ArmEmpty
del OnTable(x) ∧ ArmEmpty
add Holding(x)
```

- Example 4:

The *putdown* action occurs when the arm places the object *x* onto the table.

```
PutDown(x)
pre Holding(x)
del Holding(x)
add Holding(x) ∧ ArmEmpty
```

- What is a plan?
A sequence (list) of actions, with variables replaced by constants.
- So, to get from:



- We need the set of actions:

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

Naive Planner

- In “real life”, plans contain conditionals (IF . . . THEN . . .) and loops (WHILE . . . DO . . .), but most simple planners cannot handle such constructs — they construct *linear plans*.
- Simplest approach to planning: *means-ends analysis*.
- Involves backward chaining from goal to original state.
- Start by finding an action that has goal as post-condition. Assume this is the *last* action in plan.
- Then figure out what the previous state would have been. Try to find action that has *this* state as post-condition.
- *Recurse* until we end up (hopefully!) in original state.

```
function plan(
    d : WorldDesc,      // initial env state
    g : Goal,           // goal to be achieved
    p : Plan,           // plan so far
    A : set of actions // actions available)
1. if d ⊨ g then
2.   return p
3. else
4.   choose a in A such that
5.     add(a) ⊨ g and
6.     del(a) ⊄ g
7.   set g = pre(a)
8.   append a to p
9.   return plan(d, g, p, A)
```

- As we discussed in class, this doesn't quite specify the operation of the planner correctly.
- (Though it comes close I think).
- Line 5 tries to capture the idea that we pick a so that the result of action a will achieve some part of the goal.
- Similarly, line 6 tries to capture the idea that the items in the delete list of a are not part of the goal.
- Then line 7 tries to say that you add any preconditions of a that aren't already true to the set of things that still have to be achieved in order to satisfy the goal.

- How does this work on the previous example?

- This algorithm not guaranteed to find the plan...
- ... but it is *sound*: If it finds the plan is correct.
- Some problems:
 - negative goals;
 - maintenance goals;
 - conditionals & loops;
 - exponential search space;
 - logical consequence tests;

The Frame Problem

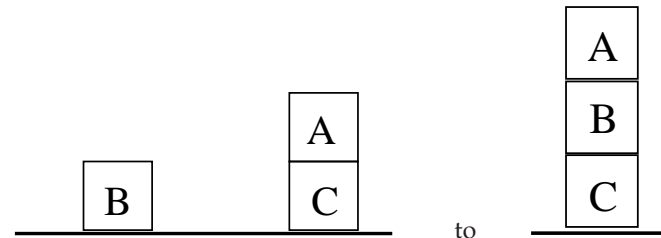
- A general problem with representing properties of actions:
 - How do we know exactly what changes as the result of performing an action?
 - If I pick up a block, does my hair colour stay the same?
- One solution is to write *frame axioms*.
 - Here is a frame axiom, which states that SP's hair colour is the same in all the situations s' that result from performing $Pickup(x)$ in situation s as it is in s .

$$\forall s, s'. Result(SP, Pickup(x), s) = s' \Rightarrow HCol(SP, s) = HCol(SP, s')$$

- Stating frame axioms in this way is unfeasible for real problems.
- (Think of all the things that we would have to state in order to cover all the possible frame axioms).
- STRIPS solves this problem by assuming that everything not explicitly stated to have changed remains unchanged.
- The price we pay for this is that we lose the advantages of using logic:
 - Semantics goes out of the window
- However, more recent work has effectively solved the frame problem (using clever second-order approaches).

Sussman's Anomaly

- Consider we have the following initial state and goal state:



- What operations will be in the plan?

- Clearly we need to *Stack* B on C at some point, and we also need to *Unstack* A from C and *Stack* it on B.
- Which operation goes first?
- Obviously we need to do the *UnStack* first, and the *Stack B* on C, but the planner has no way of knowing this.
- It also has no way of “undoing” a partial plan if it leads into a dead end.
- So if it chooses to *Stack(A, C)* after the *Unstack*, it is sunk.
- This is a big problem with linear planners
- How could we modify our planning algorithm?

- Modify the middle of the algorithm to be:

1. if $d \models g$ then
2. return p
3. else
4. choose a in A such that
5. $add(a) \models g$ and
6. $del(a) \not\models g$
- 6a. $no_clobber(add(a), del(a), rest_of_plan)$
7. set $g = pre(a)$
8. append a to p
9. return $plan(d, g, p, A)$

- But how can we do this?
- We will give an answer in the next lecture.

Summary

- This lecture has looked at planning.
- We looked mainly at a logical view of planning, using STRIPS operators.
- We also discussed the frame problem, and Sussman's anomaly.
- Sussman's anomaly motivated some thoughts about partial-order planning.
- We will cover partial order planning in more detail in the next lecture.