State-Space Planning

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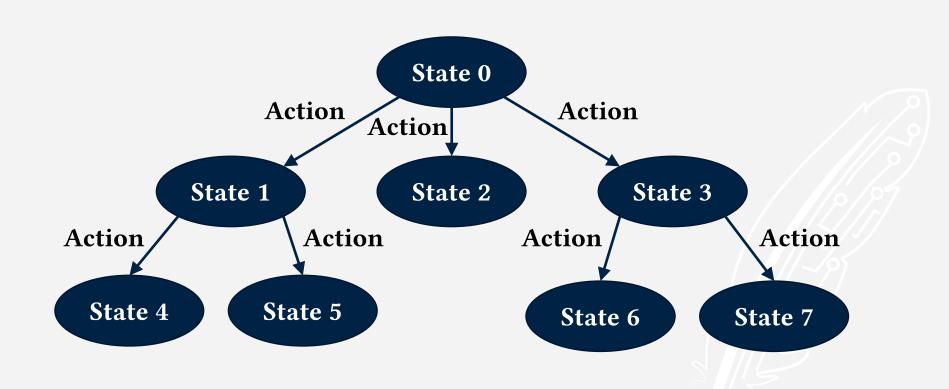


Short History of Planning

- State-space planning seems the most intuitive way to make planning into a search problem.
- However, a planning problem's state space explodes so quickly that we can't hope to search it unless we have a very accurate heuristic.
- We can reduce the number of decisions we have to make during planning through abstraction.
- Plan graphs provide an abstraction of a planning problem's search space.
- Plan graphs lead to accurate heuristics!











State-Space Planner

```
Begin with an empty priority queue.

Put the initial state onto the priority queue.

While the queue is not empty:

Pop a state C off the priority queue.

If C is a goal state, return the plan to get to C.

For every step S whose preconditions are satisfied in C:

Let N be the state after taking step S in state C.

Push N onto the queue.
```



Return failure.



A







Start by pushing the initial state onto the priority queue.

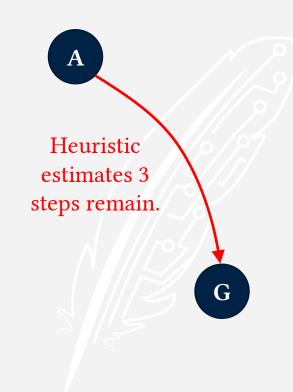
A







Start by pushing the initial state onto the priority queue.



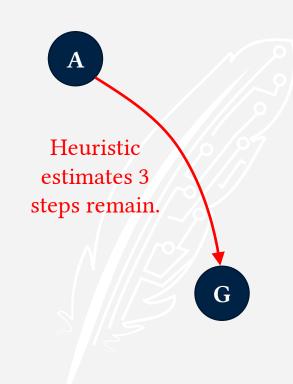




Start by pushing the initial state onto the priority queue.

Priority Queue:

A: 0 + 3 = 3







Pop a state off the queue.

Current State: A

Current Plan: Ø





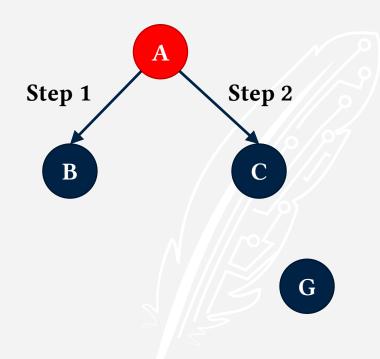




Expand the current state.

Current State: A

Current Plan: Ø



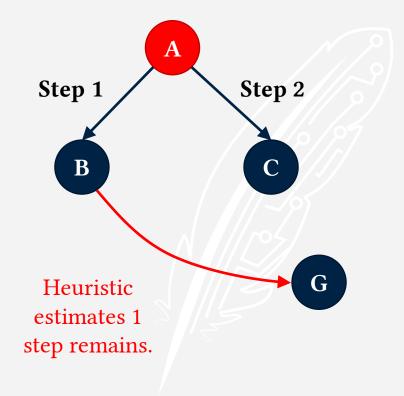




Put children on the queue.

Current State: A

Current Plan: Ø







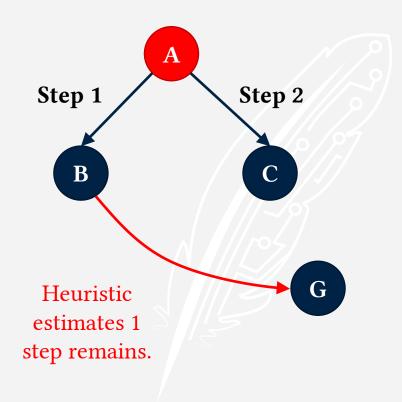
Put children on the queue.

Current State: A

Current Plan: Ø

Priority Queue:

B: 1 + 1 = 2







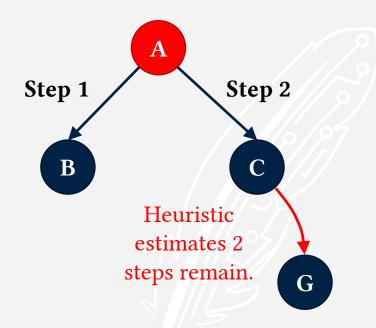
Put children on the queue.

Current State: A

Current Plan: Ø

Priority Queue:

B: 1 + 1 = 2







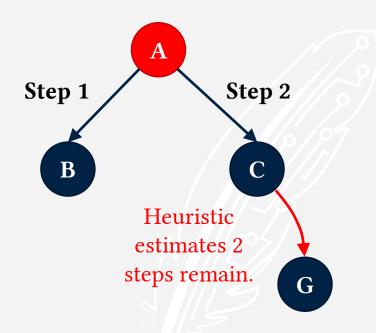
Put children on the queue.

Current State: A

Current Plan: Ø

Priority Queue:

B: 1 + 1 = 2







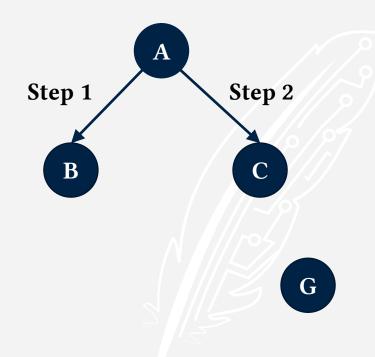
Put children on the queue.

Current State: A

Current Plan: Ø

Priority Queue:

B: 1 + 1 = 2





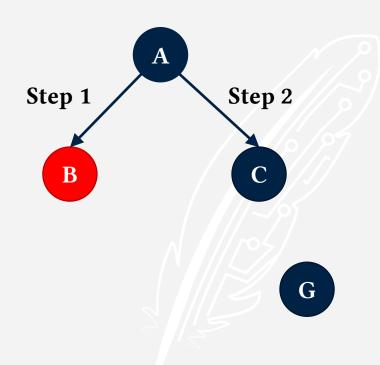


Pop a state off the queue.

Current State: B

Current Plan: 1

Priority Queue:





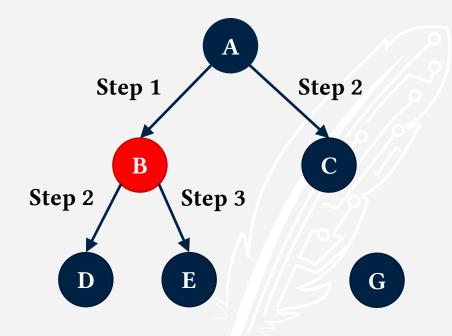


Expand the current state.

Current State: B

Current Plan: 1

Priority Queue:







Put children on the queue.

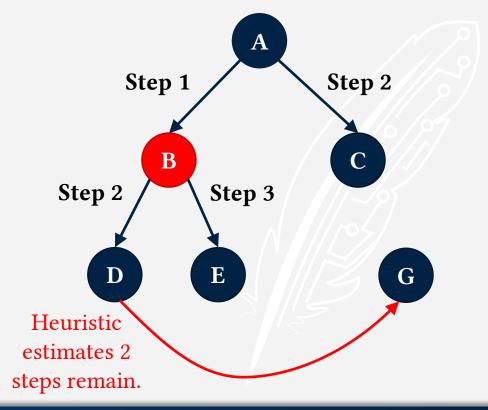
Current State: B

Current Plan: 1

Priority Queue:

C: 1 + 2 = 3

D: 2 + 2 = 4







Put children on the queue.

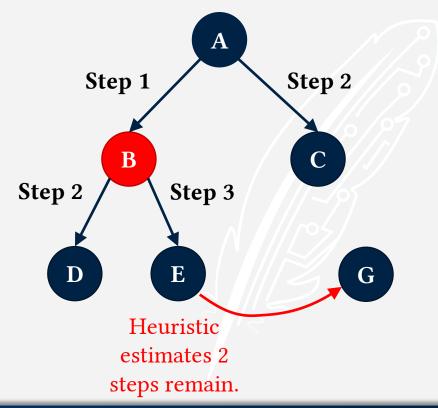
Current State: B

Current Plan: 1

$$C: 1 + 2 = 3$$

$$D: 2 + 2 = 4$$

$$E: 2 + 2 = 4$$







Put children on the queue.

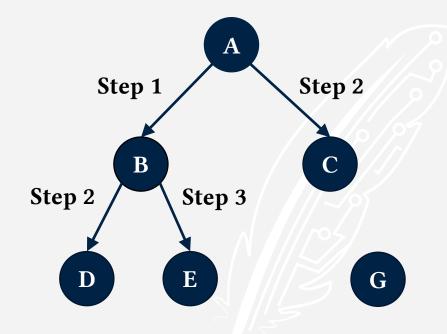
Current State: B

Current Plan: 1

$$C: 1 + 2 = 3$$

D:
$$2 + 2 = 4$$

$$E: 2 + 2 = 4$$







Pop a state off the queue.

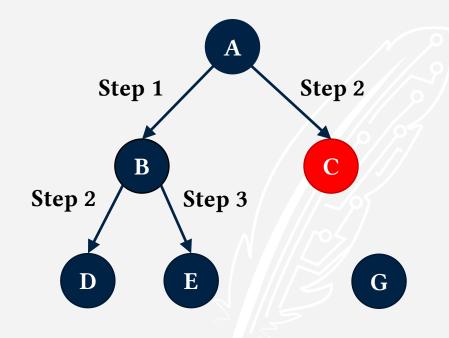
Current State: C

Current Plan: 2

Priority Queue:

D: 2 + 2 = 4

E: 2 + 2 = 4







Expand the current state.

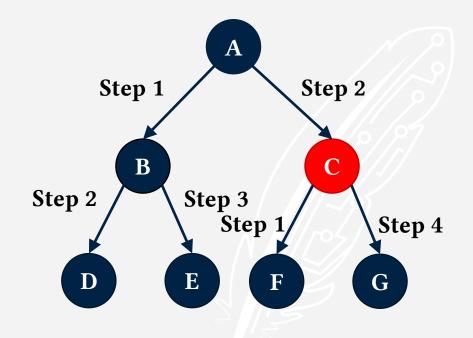
Current State: C

Current Plan: 2

Priority Queue:

D: 2 + 2 = 4

E: 2 + 2 = 4







Put children on the queue.

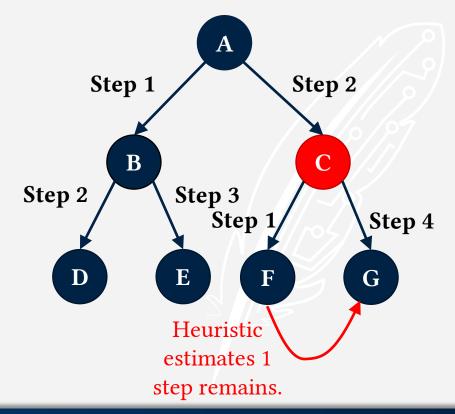
Current State: C

Current Plan: 2

$$F: 2 + 1 = 3$$

D:
$$2 + 2 = 4$$

E:
$$2 + 2 = 4$$







Put children on the queue.

Current State: C

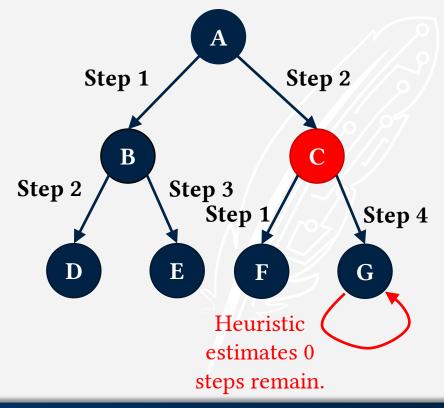
Current Plan: 2

$$G: 2 + 0 = 2$$

$$F: 2 + 1 = 3$$

D:
$$2 + 2 = 4$$

$$E: 2 + 2 = 4$$







Put children on the queue.

Current State: C

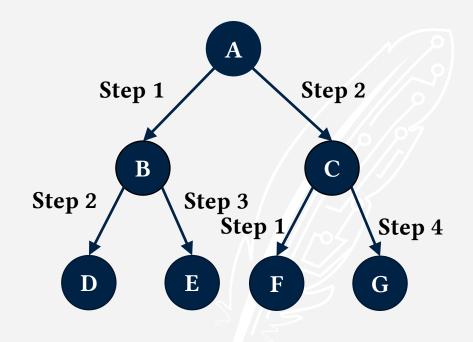
Current Plan: 2

$$G: 2 + 0 = 2$$

$$F: 2 + 1 = 3$$

D:
$$2 + 2 = 4$$

E:
$$2 + 2 = 4$$







Pop a state off the queue.

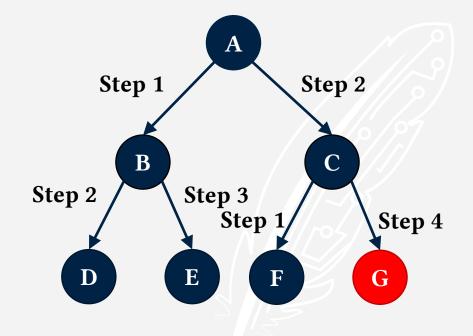
Current State: G

Current Plan: 2, 4

$$F: 2 + 1 = 3$$

D:
$$2 + 2 = 4$$

E:
$$2 + 2 = 4$$







Current state is a goal state!

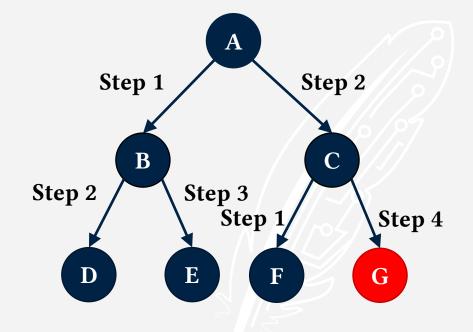
Current State: G

Current Plan: 2, 4

$$F: 2 + 1 = 3$$

D:
$$2 + 2 = 4$$

$$E: 2 + 2 = 4$$







Return plan to reach G.

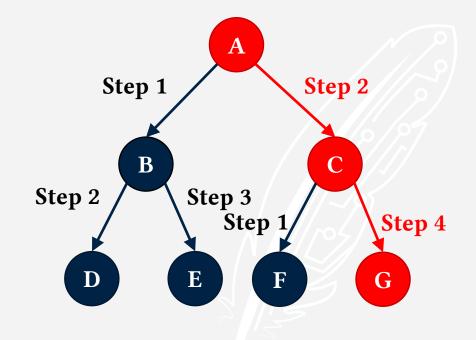
Current State: G

Current Plan: 2, 4

$$F: 2 + 1 = 3$$

$$D: 2 + 2 = 4$$

$$E: 2 + 2 = 4$$







Planning Heuristics

The speed of a state-space planner is entirely dependent on its heuristic.

State-space planning algorithms are simple; the complexity and ingenuity comes in how they calculate their heuristics.





Heuristics

A state-space planning heuristic estimates the answer to the following question: "Given some current state, how many more steps need to be taken before a goal state is reached?"

Ideally, a heuristics is:

- Highly accurate
- Admissible
- Fast to calculate





Heuristic Search Planner (HSP)

- Created by Blai Bonet, Gábor Loerincs, and Héctor Geffner
- Perhaps the first viable state-space planner.
- Winner of the first International Planning Competition in 1998





HSP's Heuristic

```
Input: The current state.
Every literal has a cost, initially ∞.
Every literal that is true in the current state has a cost of 0.
The cost of a conjunction is the sum of the costs of its conjuncts.
Do this until the costs of the literals stop changing:
    For every step S:
        For every literal E in the effect of S:
            Let the cost of E be the minimum of:
                1. The current cost of E.
                2. The cost of S's precondition + 1.
Return the cost of the problem's goal.
```

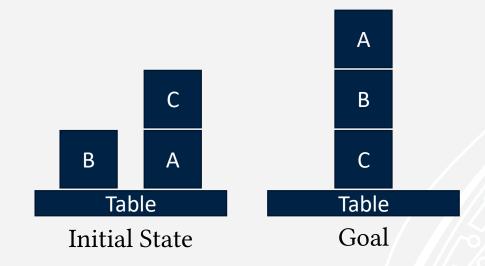




Weights:

$$h(on(A, Table)) = \infty$$

 $h(\neg on(A, Table)) = \infty$
 $h(on(A, B)) = \infty$
 $h(on(B, Table)) = \infty$
 $h(\neg on(B, Table)) = \infty$
 $h(on(B, C)) = \infty$
 $h(on(C, Table)) = \infty$
 $h(on(C, A)) = \infty$
 $h(\neg on(C, A)) = \infty$
 $h(clear(A)) = \infty$
 $h(clear(B)) = \infty$
 $h(\neg clear(C)) = \infty$
 $h(\neg clear(C)) = \infty$
 $h(\neg clear(C)) = \infty$



Note: For the sake of a small example, we will only consider *some* of the literals and actions. When HSP computes it heuristic, it considers *all* literals and *all* actions.

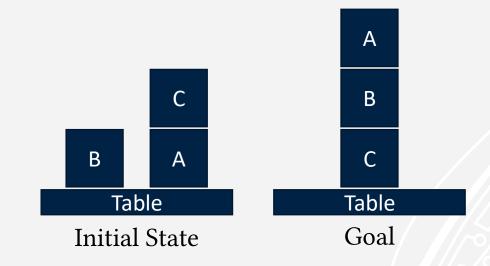




Weights:

$$h(on(A, Table)) = \infty$$

 $h(\neg on(A, Table)) = \infty$
 $h(on(A, B)) = \infty$
 $h(on(B, Table)) = \infty$
 $h(\neg on(B, Table)) = \infty$
 $h(on(B, C)) = \infty$
 $h(on(C, Table)) = \infty$
 $h(on(C, A)) = \infty$
 $h(\neg on(C, A)) = \infty$
 $h(clear(A)) = \infty$
 $h(clear(B)) = \infty$
 $h(\neg clear(C)) = \infty$
 $h(\neg clear(C)) = \infty$
 $h(\neg clear(C)) = \infty$



Start with the cost of every literal set to ∞ .

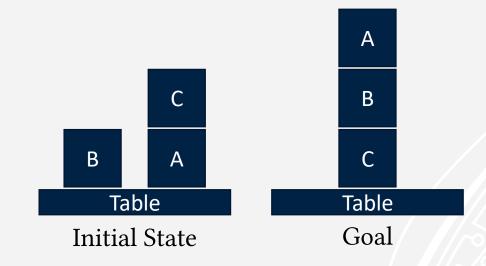




Weights:

$$h(on(A, Table)) = 0$$

 $h(\neg on(A, Table)) = \infty$
 $h(on(A, B)) = \infty$
 $h(on(B, Table)) = 0$
 $h(\neg on(B, Table)) = \infty$
 $h(on(B, C)) = \infty$
 $h(on(C, Table)) = \infty$
 $h(on(C, A)) = 0$
 $h(\neg on(C, A)) = \infty$
 $h(clear(A)) = \infty$
 $h(clear(B)) = 0$
 $h(\neg clear(C)) = 0$
 $h(\neg clear(C)) = \infty$



Set the cost of every literal that is true in the initial state to 0.

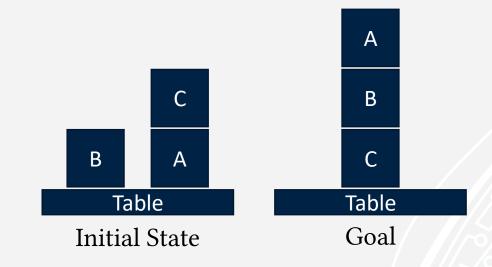




Weights:

$$h(on(A, Table)) = 0$$

 $h(\neg on(A, Table)) = \infty$
 $h(on(A, B)) = \infty$
 $h(on(B, Table)) = 0$
 $h(\neg on(B, Table)) = \infty$
 $h(on(B, C)) = \infty$
 $h(on(C, Table)) = \infty$
 $h(on(C, A)) = 0$
 $h(\neg on(C, A)) = \infty$
 $h(clear(A)) = \infty$
 $h(clear(B)) = 0$
 $h(\neg clear(C)) = 0$
 $h(\neg clear(C)) = \infty$



Update the cost of every effect of every step.





Weights:

$$h(on(A, Table)) = 0$$

$$h(\neg on(A, Table)) = \infty$$

$$h(on(A, B)) = \infty$$

$$h(on(B, Table)) = 0$$

$$h(\neg on(B, Table)) = \infty$$

$$h(on(B, C)) = \infty$$

$$h(on(C, Table)) = \infty$$

$$h(on(C, A)) = 0$$

$$h(\neg on(C, A)) = \infty$$

$$h(clear(A)) = \infty$$

$$h(clear(B)) = 0$$

$$h(\neg clear(B)) = \infty$$

$$h(clear(C)) = 0$$

$$h(\neg clear(C)) = \infty$$

$$h(clear(C)) = \infty$$

```
on(A, Table) \land clear(A) \land clear(B) move(A, Table, B) on(A, B) \land \neg on(A, Table) \land clear(Table) \land \neg clear(B)
```

Cost of precondition: $h(on(A, Table) \land clear(A) \land clear(B)) = ?$





Weights:

$$h(on(A, Table)) = 0$$

$$h(\neg on(A, Table)) = \infty$$

$$h(on(A, B)) = \infty$$

$$h(on(B, Table)) = 0$$

$$h(\neg on(B, Table)) = \infty$$

$$h(on(B, C)) = \infty$$

$$h(on(C, Table)) = \infty$$

$$h(on(C, A)) = 0$$

$$h(\neg on(C, A)) = \infty$$

$$h(clear(A)) = \infty$$

$$h(clear(B)) = 0$$

$$h(\neg clear(B)) = \infty$$

$$h(clear(C)) = 0$$

$$h(\neg clear(C)) = \infty$$

$$h(clear(C)) = \infty$$

```
on(A, Table) \land clear(A) \land clear(B) move(A, Table, B) on(A, B) \land \neg on(A, Table) \land clear(Table) \land \neg clear(B)
```

Cost of precondition: $h(on(A, Table) \land clear(A) \land clear(B)) = \infty$





Weights:

$$h(on(A, Table)) = 0$$

$$h(\neg on(A, Table)) = \infty$$

$$h(on(A, B)) = \infty$$

$$h(on(B, Table)) = 0$$

$$h(\neg on(B, Table)) = \infty$$

$$h(on(B, C)) = \infty$$

$$h(on(C, Table)) = \infty$$

$$h(on(C, A)) = 0$$

$$h(\neg on(C, A)) = \infty$$

$$h(clear(A)) = \infty$$

$$h(clear(B)) = 0$$

$$h(\neg clear(C)) = 0$$

$$h(\neg clear(C)) = \infty$$

$$h(clear(Table)) = \infty$$

$$on(A, Table) \land clear(A) \land clear(B)$$
 $move(A, Table, B)$
 $on(A, B) \land \neg on(A, Table) \land clear(Table) \land \neg clear(B)$

Cost of precondition:

$$h(on(A, Table) \land clear(A) \land clear(B)) = \infty$$

Set
$$h(on(A, B)) = min(\infty, \infty + 1)$$

Set
$$h(\neg on(A, Table)) = \min(\infty, \infty + 1)$$

Set
$$h(clear(Table)) = \min(\infty, \infty + 1)$$

Set
$$h(\neg clear(B)) = \min(\infty, \infty + 1)$$





Weights:

$$h(on(A, Table)) = 0$$

$$h(\neg on(A, Table)) = \infty$$

$$h(on(A, B)) = \infty$$

$$h(on(B, Table)) = 0$$

$$h(\neg on(B, Table)) = \infty$$

$$h(on(B, C)) = \infty$$

$$h(on(C, Table)) = \infty$$

$$h(on(C, A)) = 0$$

$$h(\neg on(C, A)) = \infty$$

$$h(clear(A)) = \infty$$

$$h(clear(B)) = 0$$

$$h(\neg clear(C)) = 0$$

$$h(\neg clear(C)) = \infty$$

$$h(clear(Table)) = \infty$$

$$on(C,A) \land clear(C)$$
 $moveToTable(C,A)$
 $on(C,Table) \land \neg on(C,A) \land clear(A)$

Cost of precondition: $h(on(C, A) \land clear(C)) = ?$





Weights:

$$h(on(A, Table)) = 0$$

$$h(\neg on(A, Table)) = \infty$$

$$h(on(A, B)) = \infty$$

$$h(on(B, Table)) = 0$$

$$h(\neg on(B, Table)) = \infty$$

$$h(on(B, C)) = \infty$$

$$h(on(C, Table)) = \infty$$

$$h(on(C, A)) = 0$$

$$h(\neg on(C, A)) = \infty$$

$$h(clear(A)) = \infty$$

$$h(clear(B)) = 0$$

$$h(\neg clear(C)) = 0$$

$$h(\neg clear(C)) = \infty$$

$$h(clear(Table)) = \infty$$

$$on(C,A) \land clear(C)$$
 $moveToTable(C,A)$
 $on(C,Table) \land \neg on(C,A) \land clear(A)$

$$h(on(C,A) \wedge clear(C)) = 0$$

Set
$$h(on(C, Table)) = min(\infty, 0 + 1)$$

Set
$$h(\neg on(C, A)) = \min(\infty, 0 + 1)$$

Set
$$h(clear(A)) = \min(\infty, 0 + 1)$$





Weights:

```
h(on(A, Table)) = 0
h(\neg on(A, Table)) = \infty
h(on(A, B)) = \infty
h(on(B, Table)) = 0
h(\neg on(B, Table)) = \infty
h(on(B,C)) = \infty
h(on(C, Table)) = 1
h(on(C,A))=0
h(\neg on(C, A)) = 1
h(clear(A)) = 1
h(clear(B)) = 0
h(\neg clear(B)) = \infty
h(clear(C)) = 0
h(\neg clear(C)) = \infty
h(clear(Table)) = \infty
```

$$on(C,A) \land clear(C)$$
 $moveToTable(C,A)$
 $on(C,Table) \land \neg on(C,A) \land clear(A)$

$$h(on(C,A) \wedge clear(C)) = 0$$

Set
$$h(on(C, Table)) = min(\infty, 0 + 1)$$

Set
$$h(\neg on(C, A)) = \min(\infty, 0 + 1)$$

Set
$$h(clear(A)) = \min(\infty, 0 + 1)$$





Weights:

```
h(on(A, Table)) = 0
h(\neg on(A, Table)) = \infty
h(on(A, B)) = \infty
h(on(B, Table)) = 0
h(\neg on(B, Table)) = \infty
h(on(B,C)) = \infty
h(on(C, Table)) = 1
h(on(C,A))=0
h(\neg on(C, A)) = 1
h(clear(A)) = 1
h(clear(B)) = 0
h(\neg clear(B)) = \infty
h(clear(C)) = 0
h(\neg clear(C)) = \infty
h(clear(Table)) = \infty
```

```
on(B, Table) \land clear(B) \land clear(C)
move(B, Table, C)
on(B, C) \land \neg on(B, Table) \land clear(Table) \land \neg clear(C)
```

Cost of precondition: $h(on(B, Table) \land clear(B) \land clear(C)) = ?$





Weights:

```
h(on(A, Table)) = 0
h(\neg on(A, Table)) = \infty
h(on(A, B)) = \infty
h(on(B, Table)) = 0
h(\neg on(B, Table)) = \infty
h(on(B,C)) = \infty
h(on(C, Table)) = 1
h(on(C,A))=0
h(\neg on(C, A)) = 1
h(clear(A)) = 1
h(clear(B)) = 0
h(\neg clear(B)) = \infty
h(clear(C)) = 0
h(\neg clear(C)) = \infty
h(clear(Table)) = \infty
```

$$on(B,Table) \land clear(B) \land clear(C)$$
 $move(B,Table,C)$
 $on(B,C) \land \neg on(B,Table) \land clear(Table) \land \neg clear(C)$

Cost of precondition:

$$h(on(B, Table) \land clear(B) \land clear(C)) = 0$$

Set
$$h(on(B,C)) = \min(\infty, 0+1)$$

Set $h(\neg on(B,Table)) = \min(\infty, 0+1)$
Set $h(clear(Table)) = \min(\infty, 0+1)$
Set $h(\neg clear(C)) = \min(\infty, 0+1)$





Weights:

```
h(on(A, Table)) = 0
h(\neg on(A, Table)) = \infty
h(on(A, B)) = \infty
h(on(B, Table)) = 0
h(\neg on(B, Table)) = 1
h(on(B,C))=1
h(on(C, Table)) = 1
h(on(C,A))=0
h(\neg on(C, A)) = 1
h(clear(A)) = 1
h(clear(B)) = 0
h(\neg clear(B)) = \infty
h(clear(C)) = 0
h(\neg clear(C)) = 1
h(clear(Table)) = 1
```

```
on(B,Table) \land clear(B) \land clear(C)
move(B,Table,C)
on(B,C) \land \neg on(B,Table) \land clear(Table) \land \neg clear(C)
```

Cost of precondition:

$$h(on(B, Table) \land clear(B) \land clear(C)) = 0$$

Set
$$h(on(B,C)) = \min(\infty, 0+1)$$

Set
$$h(\neg on(B, Table)) = \min(\infty, 0 + 1)$$

Set
$$h(clear(Table)) = \min(\infty, 0 + 1)$$

Set
$$h(\neg clear(C)) = \min(\infty, 0 + 1)$$





Weights:

$$h(on(A, Table)) = 0$$

 $h(\neg on(A, Table)) = \infty$
 $h(on(A, B)) = \infty$
 $h(on(B, Table)) = 0$
 $h(\neg on(B, Table)) = 1$
 $h(on(B, C)) = 1$
 $h(on(C, Table)) = 1$
 $h(on(C, A)) = 0$
 $h(\neg on(C, A)) = 1$
 $h(clear(A)) = 1$
 $h(clear(B)) = 0$
 $h(\neg clear(C)) = 0$
 $h(\neg clear(C)) = 1$
 $h(clear(Table)) = 1$

Some costs changed, so we need to do another round of updates.





Weights:

```
h(on(A, Table)) = 0
h(\neg on(A, Table)) = \infty
h(on(A, B)) = \infty
h(on(B, Table)) = 0
h(\neg on(B, Table)) = 1
h(on(B,C)) = 1
h(on(C, Table)) = 1
h(on(C,A))=0
h(\neg on(C, A)) = 1
h(clear(A)) = 1
h(clear(B)) = 0
h(\neg clear(B)) = \infty
h(clear(C)) = 0
h(\neg clear(C)) = 1
h(clear(Table)) = 1
```

```
on(A, Table) \land clear(A) \land clear(B)
move(A, Table, B)
on(A, B) \land \neg on(A, Table) \land clear(Table) \land \neg clear(B)
```

Cost of precondition:

$$h(on(A, Table) \land clear(A) \land clear(B)) = ?$$





Weights:

```
h(on(A, Table)) = 0
h(\neg on(A, Table)) = \infty
h(on(A, B)) = \infty
h(on(B, Table)) = 0
h(\neg on(B, Table)) = 1
h(on(B,C)) = 1
h(on(C, Table)) = 1
h(on(C,A))=0
h(\neg on(C,A)) = 1
h(clear(A)) = 1
h(clear(B)) = 0
h(\neg clear(B)) = \infty
h(clear(C)) = 0
h(\neg clear(C)) = 1
h(clear(Table)) = 1
```

$$on(A, Table) \land clear(A) \land clear(B)$$
 $move(A, Table, B)$
 $on(A, B) \land \neg on(A, Table) \land clear(Table) \land \neg clear(B)$

Cost of precondition:

$$h(on(A, Table) \land clear(A) \land clear(B)) = 1$$

Set
$$h(on(A, B)) = min(\infty, 1 + 1)$$

Set
$$h(\neg on(A, Table)) = \min(\infty, 1 + 1)$$

Set
$$h(clear(Table)) = min(1,1+1)$$

Set
$$h(\neg clear(B)) = \min(\infty, 1+1)$$





```
Weights:
```

```
h(on(A, Table)) = 0
h(\neg on(A, Table)) = 2
h(on(A,B))=2
h(on(B, Table)) = 0
h(\neg on(B, Table)) = 1
h(on(B,C)) = 1
h(on(C, Table)) = 1
h(on(C,A))=0
h(\neg on(C,A)) = 1
h(clear(A)) = 1
h(clear(B)) = 0
h(\neg clear(B)) = 2
h(clear(C)) = 0
h(\neg clear(C)) = 1
```

```
on(A, Table) \land clear(A) \land clear(B)
                    move(A, Table, B)
on(A, B) \land \neg on(A, Table) \land clear(Table) \land \neg clear(B)
```

Cost of precondition:

$$h(on(A, Table) \land clear(A) \land clear(B)) = 1$$

Set
$$h(on(A, B)) = min(\infty, 1 + 1)$$

Set
$$h(\neg on(A, Table)) = \min(\infty, 1 + 1)$$

Set
$$h(clear(Table)) = min(1,1+1)$$

Set
$$h(\neg clear(B)) = \min(\infty, 1+1)$$



h(clear(Table)) = 1



Weights:

```
h(on(A, Table)) = 0
h(\neg on(A, Table)) = 2
h(on(A,B))=2
h(on(B, Table)) = 0
h(\neg on(B, Table)) = 1
h(on(B,C)) = 1
h(on(C, Table)) = 1
h(on(C,A))=0
h(\neg on(C, A)) = 1
h(clear(A)) = 1
h(clear(B)) = 0
h(\neg clear(B)) = 2
h(clear(C)) = 0
h(\neg clear(C)) = 1
h(clear(Table)) = 1
```

```
on(C,A) \land clear(C)
moveToTable(C,A)
on(C,Table) \land \neg on(C,A) \land clear(A)
```

Cost of precondition:
$$h(on(C, A) \land clear(C)) = 0$$

Set
$$h(on(C, Table)) = min(1, 0 + 1)$$

Set $h(\neg on(C, A)) = min(1, 0 + 1)$
Set $h(clear(A)) = min(1, 0 + 1)$





Weights:

```
h(on(A, Table)) = 0
h(\neg on(A, Table)) = 2
h(on(A,B))=2
h(on(B, Table)) = 0
h(\neg on(B, Table)) = 1
h(on(B,C)) = 1
h(on(C, Table)) = 1
h(on(C,A))=0
h(\neg on(C, A)) = 1
h(clear(A)) = 1
h(clear(B)) = 0
h(\neg clear(B)) = 2
h(clear(C)) = 0
h(\neg clear(C)) = 1
h(clear(Table)) = 1
```

```
on(B, Table) \land clear(B) \land clear(C)
move(B, Table, C)
on(B, C) \land \neg on(B, Table) \land clear(Table) \land \neg clear(C)
```

$$h(on(B, Table) \land clear(B) \land clear(C)) = 0$$

$$Set h(on(B,C)) = min(1,0+1)$$

Set
$$h(\neg on(B, Table)) = \min(1, 0 + 1)$$

Set
$$h(clear(Table)) = min(1,0+1)$$

Set
$$h(\neg clear(C)) = \min(1,0+1)$$





Weights:

$$h(on(A, Table)) = 0$$

$$h(\neg on(A, Table)) = 2$$

$$h(on(A, B)) = 2$$

$$h(on(B, Table)) = 0$$

$$h(\neg on(B, Table)) = 1$$

$$h(on(B, C)) = 1$$

$$h(on(C, Table)) = 1$$

$$h(on(C, A)) = 0$$

$$h(\neg on(C, A)) = 1$$

$$h(clear(A)) = 1$$

$$h(clear(B)) = 0$$

$$h(\neg clear(C)) = 0$$

$$h(\neg clear(C)) = 1$$

$$h(clear(Table)) = 1$$

Some costs changed, so we need to do another round of updates.

Spoiler: Nothing will change this round.

After a round where no weights change, we are done.

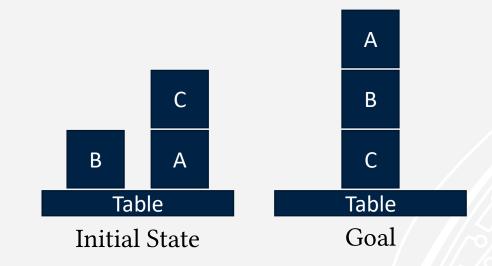




Weights:

$$h(on(A, Table)) = 0$$

 $h(\neg on(A, Table)) = 2$
 $h(on(A, B)) = 2$
 $h(on(B, Table)) = 0$
 $h(\neg on(B, Table)) = 1$
 $h(on(B, C)) = 1$
 $h(on(C, Table)) = 1$
 $h(on(C, A)) = 0$
 $h(\neg on(C, A)) = 1$
 $h(clear(A)) = 1$
 $h(clear(B)) = 0$
 $h(\neg clear(C)) = 0$
 $h(\neg clear(C)) = 1$
 $h(clear(Table)) = 1$



What is the estimated cost of the goal? $h(on(A, B) \land on(B, C)) = ?$

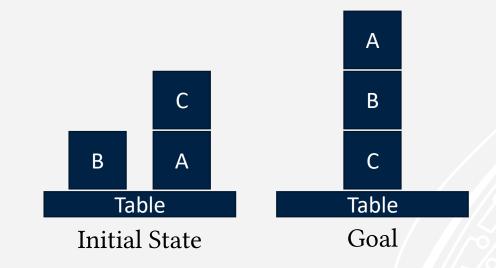




Weights:

$$h(on(A, Table)) = 0$$

 $h(\neg on(A, Table)) = 2$
 $h(on(A, B)) = 2$
 $h(on(B, Table)) = 0$
 $h(\neg on(B, Table)) = 1$
 $h(on(B, C)) = 1$
 $h(on(C, Table)) = 1$
 $h(on(C, A)) = 0$
 $h(\neg on(C, A)) = 1$
 $h(clear(A)) = 1$
 $h(clear(B)) = 0$
 $h(\neg clear(C)) = 0$
 $h(\neg clear(C)) = 1$
 $h(clear(Table)) = 1$



What is the estimated cost of the goal? $h(on(A, B) \land on(B, C)) = 2 + 1 = 3$





HSP's Heuristic





Planning Problems

During planning, goals can:

- Interfere: Progress toward one goal undoes progress toward another goal.
- **Synergize:** Progress toward one goal also makes progress toward another goal.





HSP's Heuristic

• HSP does not account for interference.

(Sometimes called the "ignore delete list" assumption)

HSP does not account for synergy.

(Because the cost of a goal is the sum of its parts)

- HSP may overestimate, and thus is not admissible.
- In practice, HSP was the first heuristic that was accurate enough to allow for state-space planning.
- Heuristic is efficient to compute.





Analyzing HSP's Heuristic

- The heuristic is based on a relaxed version of the planning problem that is much easier to solve but still provides a good approximation of the original.
- Relaxed problem: We can never get farther from a goal, only closer (i.e. ignore interference).
- The cost of achieving some literal is 1 + the cost of achieving the precondition of any step which has that literal as an effect.





Fast-Forward (FF)

- Created by Bernhard Nebel and Jörg Hoffmann
- Top performer in the second (2000) and third (2002) International Planning Competitions
- Observation: HSP's estimates are very similar to those obtained from a plan graph
- Idea: Use a plan graph to estimate the difficulty of a goal and to find a solution to the relaxed problem
- Benefit: Plan graphs account for synergy in goals





FF Heuristic

```
Given a current state S and a goal G:

Construct a plan graph such that layer 0 is S.

Extend the plan graph until all goals in G appear.

(Do not use persistence steps or mutexes.)

Extract a solution as graphplan does.

Return the size of the resulting plan.
```





Initial State: $at(C1, ATL) \land at(C2, ATL) \land at(P1, ATL)$

Goal: $at(C1, MSY) \land at(C2, MSY)$

Plan:

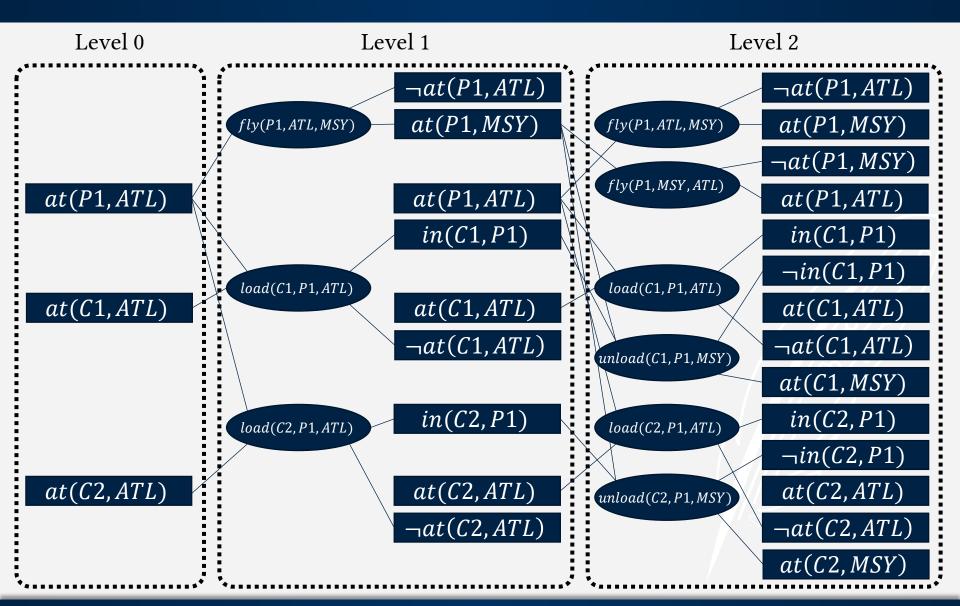
- 1. load(C1, P1, ATL)
- 2. load(C2, P1, ATL)
- 3. fly(P1, ATL, MSY)
- 4. unload(C1, P1, MSY)
- $5. \quad unload(C2, P1, MSY)$

HSP Estimate: ? steps

Plan Graph Estimate: ? steps

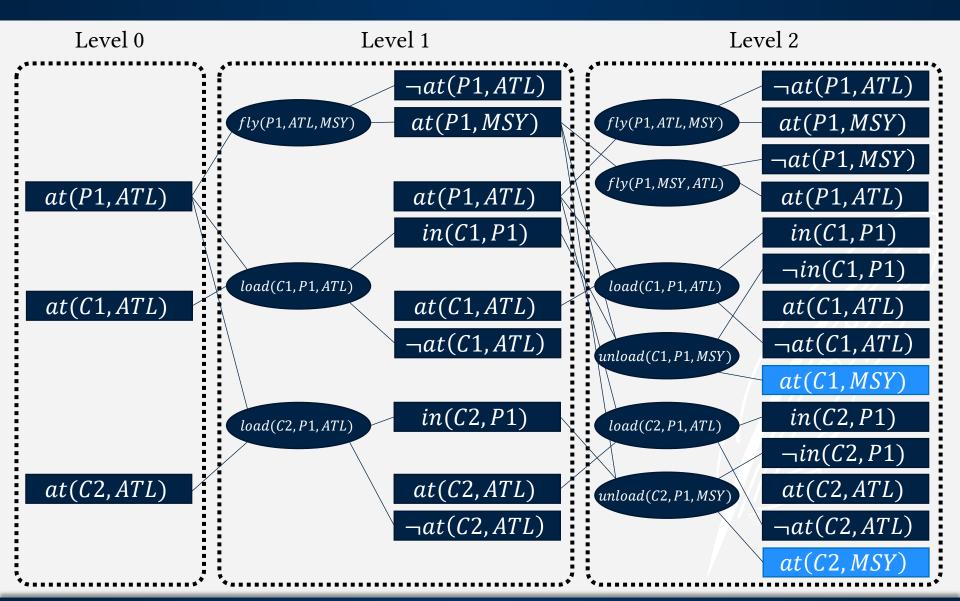






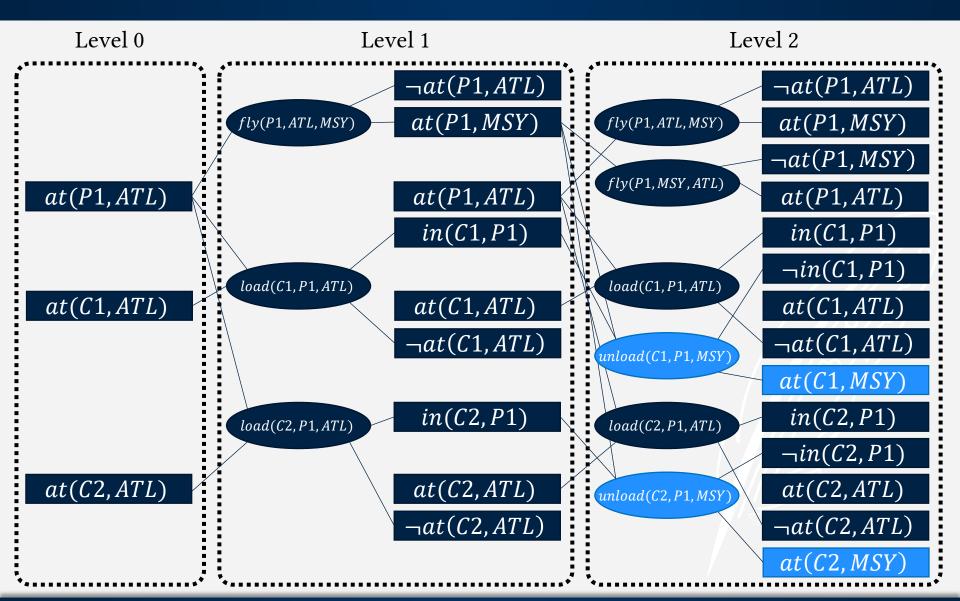






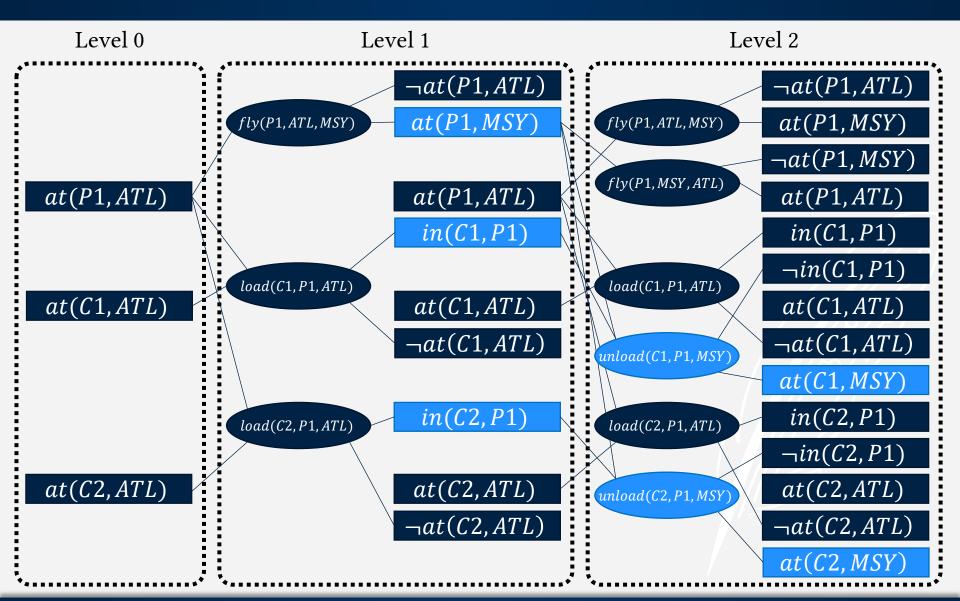






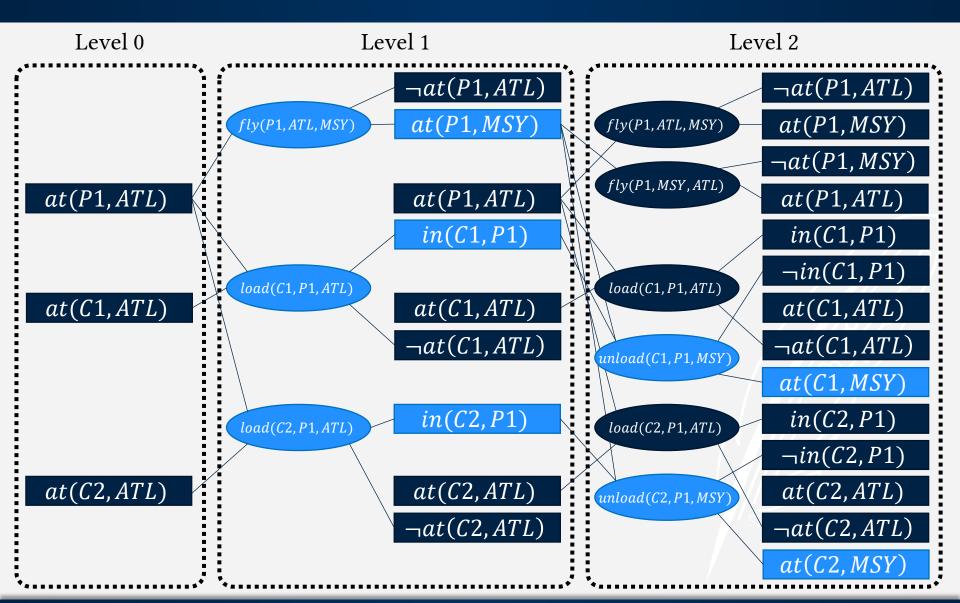






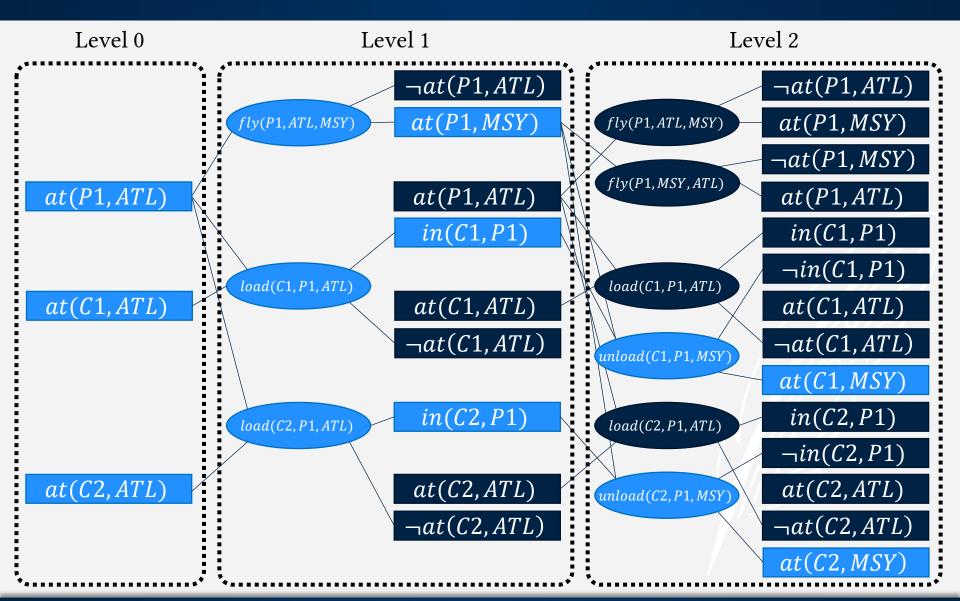
















Initial State: $at(C1, ATL) \land at(C2, ATL) \land at(P1, ATL)$

Goal: $at(C1, MSY) \land at(C2, MSY)$

Plan:

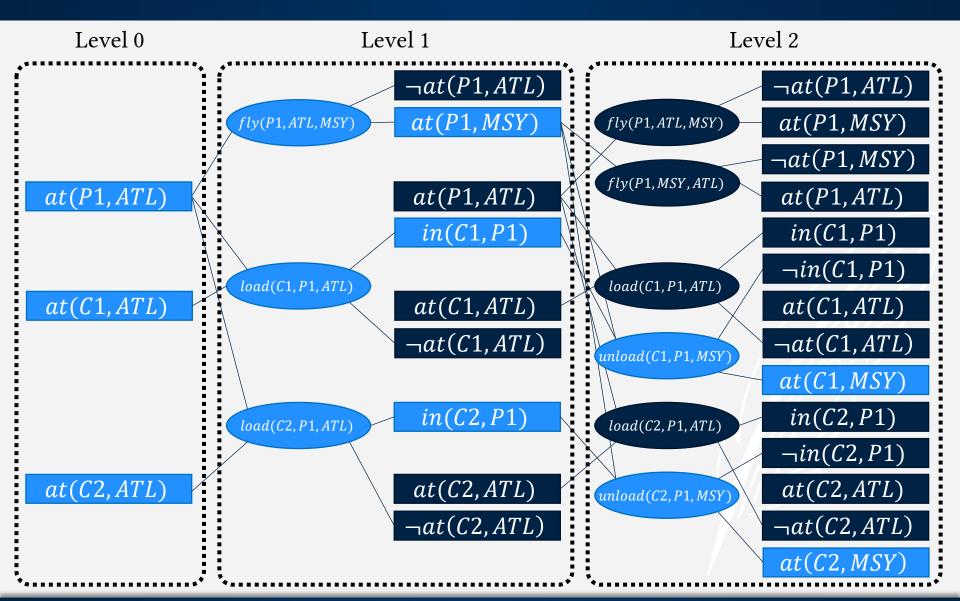
- 1. load(C1, P1, ATL)
- 2. load(C2, P1, ATL)
- 3. fly(P1, ATL, MSY)
- 4. unload(C1, P1, MSY)
- $5. \quad unload(C2, P1, MSY)$

HSP Estimate: ? steps

Plan Graph Estimate: ? steps











Initial State: $at(C1, ATL) \land at(C2, ATL) \land at(P1, ATL)$

Goal: $at(C1, MSY) \land at(C2, MSY)$

Plan:

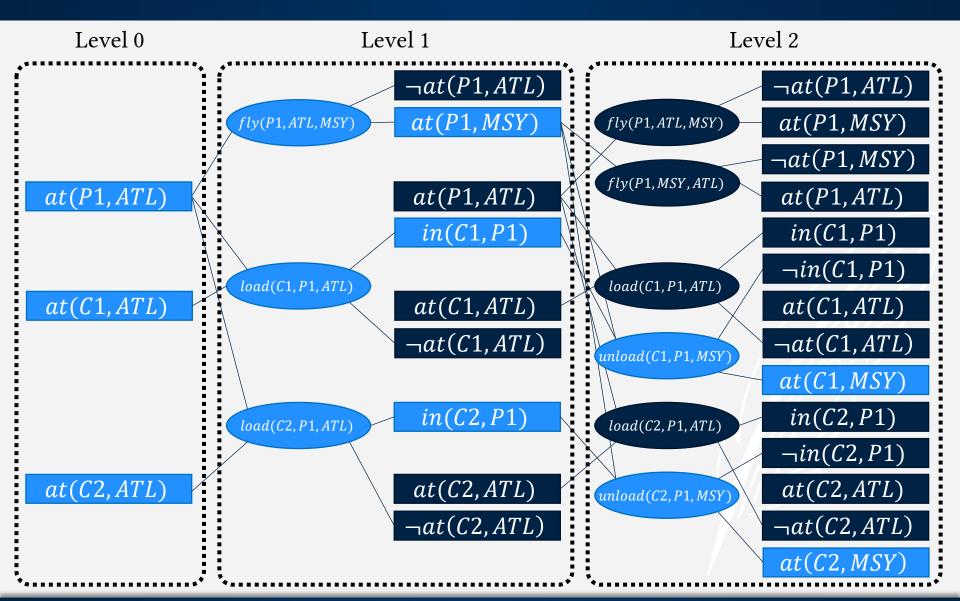
- 1. load(C1, P1, ATL)
- 2. load(C2, P1, ATL)
- 3. fly(P1, ATL, MSY)
- 4. unload(C1, P1, MSY)
- $5. \quad unload(C2, P1, MSY)$

HSP Estimate: 6 steps

Plan Graph Estimate: ? steps











Initial State: $at(C1, ATL) \land at(C2, ATL) \land at(P1, ATL)$

Goal: $at(C1, MSY) \land at(C2, MSY)$

Plan:

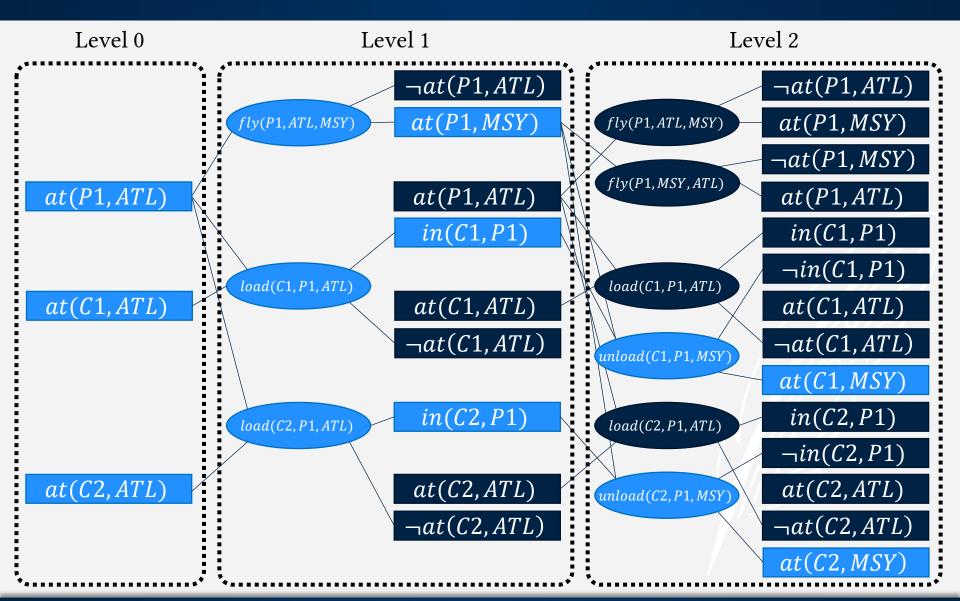
- 1. load(C1, P1, ATL)
- 2. load(C2, P1, ATL)
- 3. fly(P1, ATL, MSY)
- 4. unload(C1, P1, MSY)
- $5. \quad unload(C2, P1, MSY)$

HSP Estimate: 6 steps

Plan Graph Estimate: 2 steps











Cargo Problem

Initial State: $at(C1, ATL) \land at(C2, ATL) \land at(P1, ATL)$

Goal: $at(C1, MSY) \land at(C2, MSY)$

Plan:

- 1. load(C1, P1, ATL)
- 2. load(C2, P1, ATL)
- 3. fly(P1, ATL, MSY)
- 4. unload(C1, P1, MSY)
- $5. \quad unload(C2, P1, MSY)$

HSP Estimate: 6 steps

Plan Graph Estimate: 2 steps

FF Estimate: 5 steps





Graphplan vs. HSP vs. FF

- All three are solving a relaxed version of the problem by never deleting facts.
- Simply using the level of the plan graph at which the goal first appears is admissible but often underestimates dramatically.
- HSP does not account for synergy between goals and so is more prone to overestimate.
- FF accounts for synergy between goals and so often gives more accurate estimates.





Mutexes

FF does not compute mutexes when extending the plan graph. This is for two reasons:

- In practice, the extra accuracy gained by using mutexes is not worth the cost of computing them.
- Graphplan solution extraction is P-SPACE-hard because of mutexes. FF must re-compute the plan graph at every iteration of the search, so extracting a solution with mutexes is way too expensive. Without mutexes, solution extraction can be done greedily and is only P-TIME-hard.





Fast-Downward

- Created by Malte Helmert and Silvia Richter
- Winner of the fourth International Planning Competition (2004)
- Every winning planner since then up until the present has been based on FD





Fast-Downward

- Translates propositional problem representation into a variable / value representation, which allows for smaller and faster data structures.
- Computes domain transition graph to describe how a variable's value can change.
- Relaxed plans extracted from domain transition graphs.





Propositional Representation

Given 1 plane and 4 airports, there are 8 literals the need to be expessed:

- 1. at(P1, ATL)
- 2. $\neg at(P1, ATL)$
- 3. at(P1, MSY)
- 4. $\neg at(P1, MSY)$
- 5. at(P1, SFO)
- 6. $\neg at(P1, SF0)$
- 7. at(P1, DFW)
- 8. $\neg at(P1, DFW)$

Traditionally, a state has been represented as an array of Boolean variables. Given a problem with *p* planes and *a* airports, how many indices are needed in this array?

$$p \cdot a$$





Variable / Value Representation

Rather than representing literals as Boolean values, FD infers a set of variables which can have one of many possible values:

- 1. location(P1) = ATL
- 2. location(P1) = MSY
- 3. location(P1) = SFO
- 4. location(P1) = DFW

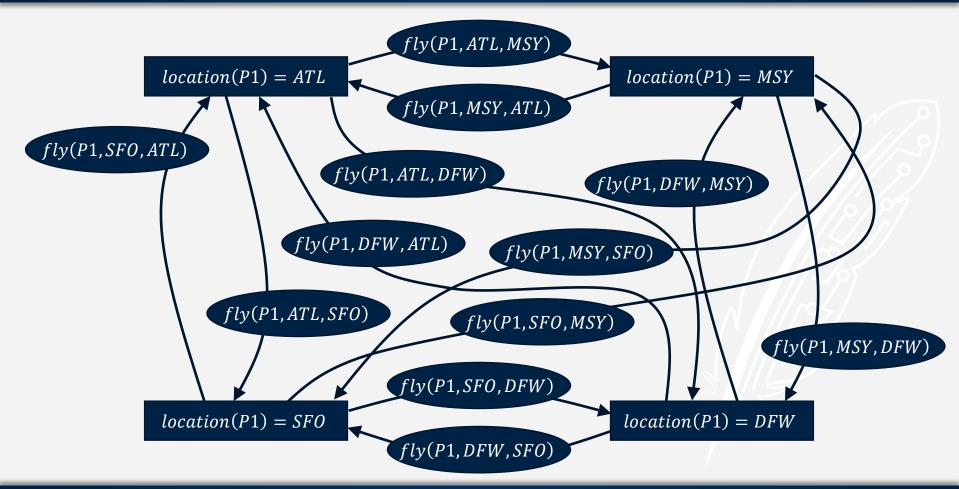
Given a problem with *p* planes and *a* airports, how many indices are needed in this new array?

p





Domain Transition Graphs







FD Heuristic (Overview)

- FD calculates it heuristic by considering domain transition graphs (DTGs).
- Given the current value of a variable (current node in the DTG), follow edges (steps) until we reach the value of that variable in the goal.
- The order in which variables are considered is decided based on how goals interact.
- FD is essentially breaking down each goal into a causal chain (remember causal links?).





HSP vs. FF vs. FD

All three planners calculate their heuristics by solving a relaxed version of the problem that is only P-TIME-hard instead of P-SPACE-hard:

- HSP does not account for synergy or interference.
- FF accounts for synergy but not interference.
- FD accounts for synergy and some interference.





History of Planning

- State space planning would be the most straightforward way to approach the problem, but the search space explodes so quickly that this is only viable with highly accurate heuristics.
- Other approaches to planning (e.g. POCL and Graphplan) are developed. They use abstraction to reduce the number of decisions the planner makes.





History of Planning

- Plan graphs leads to the development of accurate heuristics.
- Heuristics become progressively more accurate by solving relaxed problems which are more and more similar to the original problem.
- Other ideas developed early in the history of planning research (e.g. causal links) are often helpful in developing new approaches.



