

AIND Planning Assignment

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1 Problem Definition

Given three problems in the Air Cargo domain, we are asked to provide a planning search algorithm to find an optimal plan for solving these problems. The Air Cargo Problem Action Schema is as follows.

Action(*Load*(*c*, *p*, *a*),
PRECOND : $At(c, a) \wedge At(p, a) \wedge Cargo(c) \wedge Plane(p) \wedge Airport(a)$,
EFFECT : $\neg At(c, a) \wedge In(c, p)$)
Action(*Unload*(*c*, *p*, *a*),
PRECOND : $In(c, p) \wedge At(p, a) \wedge Cargo(c) \wedge Plane(p) \wedge Airport(a)$,
EFFECT : $At(c, a) \wedge \neg In(c, p)$)
Action(*Fly*(*p*, *from*, *to*),
PRECOND : $At(p, from) \wedge Plane(p) \wedge Airport(from) \wedge Airport(to)$,
EFFECT : $\neg At(p, from) \wedge At(p, to)$)

1.0.1 Problem 1

Init State:

$(At(C1, SFO) \wedge At(C2, JFK)$
 $\wedge At(P1, SFO) \wedge At(P2, JFK)$
 $\wedge Cargo(C1) \wedge Cargo(C2)$
 $\wedge Plane(P1) \wedge Plane(P2)$
 $\wedge Airport(JFK) \wedge Airport(SFO))$

Goal State:

$(At(C1, JFK) \wedge At(C2, SFO))$

Optimal Plan:

Load(C1, P1, SFO)
Load(C2, P2, JFK)
Fly(P1, SFO, JFK)
Fly(P2, JFK, SFO)
Unload(C1, P1, JFK)

Unload(C2, P2, SFO)

1.0.2 Problem 2

Init State:

$Init(At(C1, SFO) \wedge At(C2, JFK) \wedge At(C3, ATL)$
 $\wedge At(P1, SFO) \wedge At(P2, JFK) \wedge At(P3, ATL)$
 $\wedge Cargo(C1) \wedge Cargo(C2) \wedge Cargo(C3)$
 $\wedge Plane(P1) \wedge Plane(P2) \wedge Plane(P3)$
 $\wedge Airport(JFK) \wedge Airport(SFO) \wedge Airport(ATL))$

Goal State:

$(At(C1, JFK) \wedge At(C2, SFO) \wedge At(C3, SFO))$

Optimal Plan:

Load(C1, P1, SFO)
Load(C2, P2, JFK)
Load(C3, P3, ATL)
Fly(P1, SFO, JFK)
Fly(P2, JFK, SFO)
Fly(P3, ATL, SFO)
Unload(C1, P1, JFK)
Unload(C2, P2, SFO)
Unload(C3, P3, SFO)

1.0.3 Problem 3

Init State:

$(At(C1, SFO) \wedge At(C2, JFK) \wedge At(C3, ATL) \wedge At(C4, ORD)$
 $\wedge At(P1, SFO) \wedge At(P2, JFK)$
 $\wedge Cargo(C1) \wedge Cargo(C2) \wedge Cargo(C3) \wedge Cargo(C4) \wedge Plane(P1) \wedge Plane(P2)$
 $\wedge Airport(JFK) \wedge Airport(SFO) \wedge Airport(ATL) \wedge Airport(ORD))$

Goal State:

$(At(C1, JFK) \wedge At(C3, JFK) \wedge At(C2, SFO) \wedge At(C4, SFO))$

Optimal Plan:

Load(C1, P1, SFO)
Load(C2, P2, JFK)
Fly(P1, SFO, ATL)
Fly(P2, JFK, ORD)
Load(C3, P1, ATL)
Load(C4, P2, ORD)
Fly(P1, ATL, JFK)

Fly(P2, ORD, SFO)
 Unload(C1, P1, JFK)
 Unload(C3, P1, JFK)
 Unload(C2, P2, SFO)
 Unload(C4, P2, SFO)

2 Benchmarking Setup

All the search algorithms were run on a machine with the following configuration

- Ubuntu 14.04 x86_64 architecture
- 128 GB GDDR3 RAM
- 40 cores of Intel(R) Xeon(R) CPU E5-2690 v2 @ 3.00GHz
- PyPy3.5

3 Non-Heuristic Methods

Problem 1

| Algorithm | Optimal | Expansions | Goal Tests | New Nodes | Time |
|---------------|---------|------------|------------|-----------|------|
| Breadth First | Yes | 43 | 56 | 180 | 0.34 |
| Depth First | No | 12 | 13 | 48 | 0.13 |
| Uniform Cost | Yes | 55 | 57 | 224 | 0.35 |

Problem 2

| Algorithm | Optimal | Expansions | Goal Tests | New Nodes | Time |
|---------------|---------|------------|------------|-----------|-------|
| Breadth First | Yes | 3343 | 4609 | 30509 | 62.09 |
| Depth First | No | 582 | 583 | 5211 | 9.75 |
| Uniform Cost | Yes | 4852 | 4854 | 44030 | 86.52 |

Problem 3

| Algorithm | Optimal | Expansions | Goal Tests | New Nodes | Time |
|---------------|---------|------------|------------|-----------|--------|
| Breadth First | Yes | 14663 | 18098 | 129631 | 312.14 |
| Depth First | No | 627 | 628 | 5176 | 11.84 |
| Uniform Cost | Yes | 18235 | 18237 | 159716 | 374.37 |

For all three problems, depth first search didn't find the optimal plan. But at the same time, it was the fastest search algorithm with the fewest expansions, goal tests and new nodes. It requires lesser memory as well to execute. For bigger problem state spaces, it performs poorly. The solution provided by depth first search is farther from the optimal solution as the problem state space grows.

Breadth first search provides the optimal plan for all three problems. It takes more time and memory than depth first search but requires less time and memory than uniform cost search.

Uniform cost search also provides the optimal plan for all three problems. It actually does the most number of expansions and goal tests. It is memory intensive compared to other algorithms since it uses a priority queue to compute the path with shortest cost to reach the goal.

Depth first search should be used whenever the agent has constrained resources for computation and finding an optimal solution is not important and any sub-optimal solution which is off by a factor of 2 to 3 works. Breadth first search should be used whenever optimal solution is required since it is more time and space efficient compared to uniform cost search. Even though BFS is complete and optimal, number of nodes expanded approaches $O(b^d)$ where b is branching factor and d is depth of tree. Thus it is not suitable for problems with high branching factor.

3.1 Heuristic Methods

Problem 1

| Algorithm | Optimal | Expansions | Goal Tests | New Nodes | Time |
|------------------------|---------|------------|------------|-----------|------|
| A^*_H1 | Yes | 55 | 57 | 224 | 0.34 |
| $A^*_H.Ignore.Precond$ | Yes | 41 | 43 | 170 | 0.86 |
| $A^*_H.PG.Level.Sum$ | Yes | 11 | 13 | 50 | 2.04 |

Problem 2

| Algorithm | Optimal | Expansions | Goal Tests | New Nodes | Time |
|------------------------|---------|------------|------------|-----------|--------|
| A^*_H1 | Yes | 4852 | 4854 | 44030 | 84.57 |
| $A^*_H.Ignore.Precond$ | Yes | 1450 | 1452 | 13303 | 64.24 |
| $A^*_H.PG.Level.Sum$ | Yes | 86 | 88 | 841 | 534.52 |

Problem 3

| Algorithm | Optimal | Expansions | Goal Tests | New Nodes | Time |
|------------------------|---------|------------|------------|-----------|--------|
| A^*_H1 | Yes | 18235 | 18237 | 159716 | 365.99 |
| $A^*_H.Ignore.Precond$ | Yes | 5040 | 5042 | 44944 | 229.15 |
| $A^*_H.PG.Level.Sum$ | Yes | 86 | 88 | 841 | NA |

A^* algorithm was tested with three heuristic functions $H1$, $H.Ignore.Preconditions$ and $H.PG.Level.Sum$. All three methods performed very well and found the optimal plan for all the problems.

As evidenced by the data presented in the tables, A^* with Level Sum heuristic performs the fewest number of expansions and goal tests. It is the heuristic function that requires least amount of memory. At the same time, Level Sum heuristic takes the maximum time to execute. This was because every time the

routine for heuristic ran, the planning graph had to be constructed. For problem 3, the program ran for more than 30 minutes and had to be terminated.

A^* with Ignore Preconditions heuristic took the least amount of time to execute. A^* with H1 heuristic was the most memory intensive and did the most number of expansions and goal tests. This wasn't surprising because H1 was a dummy heuristic which doesn't approximate the true distance to the goal correctly. Moreover, it is not admissible. Because of this, the search might be lead astray. Generally, A^* should perform better than BFS even with any reasonable admissible heuristic since it is goal directed. The Ignore Preconditions heuristic is a very good estimator of distance to goal since it approximately computes the number of actions discarding their preconditions that need to be done to meet the conditions of the goal state. It will always underestimate the true distance to the goal state ie. it's admissible and hence it is guaranteed to find the optimal solution.

4 Best Search Algorithm

Considering the speed of execution, optimality of solution, memory requirements and goal tests for each search algorithm, the best algorithm for solving the air cargo problems is A^* with Ignore Preconditions heuristic. It compares favourably with uninformed non-heuristic search algorithms and beats the best performing Breadth First Search algorithm by quite a margin. It requires less memory but performs much faster than breadth first search. Refer the graph attached below to look at the run time comparison for all six search algorithms analysed so far (both heuristic and non-heuristic).

