

Heuristic Analysis

Problem 1

Optimal plan for problem 1

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

Search method comparison for problem 1

	Optimality	Plan Length	Time Elapsed	Expansions	Goal Tests	New Nodes
breadth_first_search	yes	6	0.023	43	56	180
depth_first_graph_search	no	20	0.010	21	22	84
uniform_cost_search	yes	6	0.028	55	57	224
greedy_best_first_graph_search with h_1	yes	6	0.003	7	9	28
a* with h_1	yes	6	0.030	55	57	224
a* with h_ignore_preconditions	yes	6	0.033	41	43	170
a* with h_pg_levelsum	yes	6	1.525	11	13	50
a* with h_pg_setlevel	yes	6	1.117	8	10	35

Observation and analysis for problem 1

- *greedy_best_first_graph_search* outperform every other methods in every metrics, and accidentally every nodes popped in this search method lead to the optimal plan(just expand 7 nodes), it is because:
 - This method was implemented by *best_first_graph_search* with *PriorityQueue* where $f=1$:

```
greedy_best_first_graph_search = best_first_graph_search(problem, lambda node: 1)
```
 - With the same priority '1', 'the smaller state' of a node will be pop first, because in Node class:

```
class Node:
    def __lt__(self, node):
        return self.state < node.state
```

- The state of a node in these problems was like 'FFTFFFTFTFFF', so it intend to pop a state, with more initial 'F', first. Possitive sentences stand in front of a state, so it intent to expands the nodes, which turn possitive sentence of a state to negative, first. Which leads to the goal in these problems.
- The order of the expanded nodes is defined by `get_actions` which implemented by `return load_actions() + unload_actions() + fly_actions()`
This order is similar as the order of a optimal plan 'Load -> Fly -> Unload'. ('Load -> Unload' will rarely happend in a search because it would generate a explored node)
- Only 2 cargos and 2 planes and 2 airports, so very possible actions in order 'Load -> Fly -> Unload' can leads to the optimal plan.
- *depth_first_graph_search* always expands the deepest node, so this method is not optimal.(Russell-Norvig AIMA 3rd Ed 3.4.3) It use a LIFO stack to implement the frontier, so it pop nodes, in reverse of the actions('Load -> Unload -> Fly'), in this order 'Fly -> Unload -> 'Load'. It's a bit far away from the optimal model, so this method take more steps than *greedy_best_first_graph_search*. But it also faster than many other methods, for it just go deeper to find a solution and it expand less nodes.
- Astar is optimally efficient ofr any given consistent heuristic.(Russell-Norvig AIMA 3rd Ed 3.5.2)
- *astar_with_h_pg_levelsum* and *astar_with_h_pg_setlevel* expand less nodes than other methods which shows they have good heuristic, but too much time they take shows they were calculate too slow for the complexity of planning graph.
- This problem has 12 fluents, so the search space has $2^{12} = 4096$ states, which is not a big number so the time elapsed of every methods in this problem is not significantly different.

Problem 2

Optimal plan for problem 2

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

Search method comparison for problem 2

	Optimality	Plan Length	Time Elapsed	Expansions	Goal Tests	New Nodes
breadth_first_search	yes	9	10.284	3343	4609	30509
depth_first_graph_search	no	619	2.598	624	625	5602
uniform_cost_search	yes	9	32.155	4853	4855	44041
greedy_best_first_graph_search with h_1	no	21	5.227	998	1000	8982
a* with h_1	yes	9	32.655	4853	4855	44041
a* with h_ignore_preconditions	yes	9	10.541	1506	1508	13820
a* with h_pg_levelsum	yes	9	176.182	86	88	841
a* with h_pg_setlevel	yes	9	160.780	89	91	778

Observation and analysis for problem 2

- *depth_first_graph_search* is the fast solution but not optimal.
- *breadth_first_search* is the fastest solution in optimal methods. It is always a optimal search (Russell-Norvig AIMA 3rd Ed 3.4.1).
- *uniform_cost_search* is also a optimal search (Russell-Norvig AIMA 3rd Ed 3.4.2). But it take more time than *breadth_first_search* because *breadth_first_search* immediately stop the search after expanding the goal node but *uniform_cost_search* do the goal_test when pop the node later.
- *astar_with_h_ignore_preconditions* is the fastest solution in astar methods and not too much slower than *breadth_first_search*
- *astar_with_h_pg_levelsum* and *astar_with_h_pg_setlevel* also perform well in node expansions
- This problem has 27 fluents, so the search space has $2^{27} = 134217728$ states, which is significantly larger than problem 1. Some methods like *breadth_first_tree_search* cannot get a solution in reasonable time.

Problem 3

Optimal plan for problem 3

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

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

Search method comparison for problem 3

	Optimality	Plan Length	Time Elapsed	Expansions	Goal Tests	New Nodes
breadth_first_search	yes	12	74.178	14663	18098	129631
depth_first_graph_search	no	392	1.305	408	409	3364
uniform_cost_search	yes	12	273.602	18223	18225	159618
greedy_best_first_graph_search with h_1	no	22	71.990	5578	5580	49150
a* with h_1	yes	12	274.343	18223	18225	159618
a* with h_ignore_preconditions	yes	12	63.538	5118	5120	45650
a* with h_pg_levelsum	yes	12	1187.516	414	416	3818
a* with h_pg_setlevel	no	13	942.667	367	369	3178

Observation and analysis for problem 3

- *greedy_best_first_graph_search with h_1* works not well as before because the optimal model in this problem is no longer 'Load -> Fly -> Unload'. In this problem the optimal model is 'Load -> Fly -> Load -> Fly -> Unload', so this method lose it's innate advantage.
- *depth_first_graph_search* is the fast solution but not opimal.
- *breadth_first_search* is no longer the fastest solution in optimal methods for more bigger search space. But simple enough so it's not a bad solution.
- *uniform_cost_search* and *astar_with_h_1* has same result in very problems which is not in accident, because implementation of these two methods:
 - `uniform_cost_search = best_first_graph_search(problem, lambda node: node.path_cost)`
 - `astar_with_h_1 = best_first_graph_search(problem, lambda node: node.path_cost + 1)`
- *astar_with_h_ignore_preconditions* is the fastest solution in every methods. It should caused by the simple implementation of the heuristic in this large search space.
- *astar_with_h_pg_levelsum* and *astar_with_h_pg_setlevel* also perform well in node expansions. But the former

give a optimal plan and the latter don't. This is strange for the former can be inadmissible (but works well in practice for problems that are largely decomposable) and the latter is admissible (Russell-Norvig AIMA 3rd Ed 3.10.3). I will debug deeper later for this potential bug.

- This problem has 32 fluents, so the search space has $2^{32} = 4294967296$ states. Every methods take more time and spaces to get a plan.

Conclusions

As shown above, for a optimal plan:

- In comparison of time elapsed, *astar_with_h_ignore_preconditions* is better when search spaces is large and *breadth_first_search* is better when search spaces is small.
- In comparison of node expansions, *astar_with_h_pg_levelsum* is better.