

Planning Problem Representation

One of the optimal sequences for each problem is below. Some parts might be different a little among each algorithm. (For example, the Load orders of C1 and C2 might be different in Problem 1.) This sample was derived from breadth_first_search.

Problem1

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

Problem2

- Load(C2, P2, JFK)
- Load(C1, P1, SFO)
- 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)

Problem3

- Load(C2, P2, JFK)
- Load(C1, P1, SFO)
- 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)

Performance Comparison

non-heuristic search

Problem	Algorithm	Path Length	Expansions	Elapsed Time (sec)
Problem1	breadth-first	6	43	0.150
Problem1	depth-first	12	12	0.037
Problem1	uniform-cost	6	55	0.182
Problem2	breadth-first	9	3343	69
Problem2	depth-first	582	582	11.4
Problem2	uniform-cost	9	4852	123.6
Problem3	breadth-first	12	14663	475
Problem3	depth-first	596	627	15.1
Problem3	uniform-cost	12	18235	1255

- about problems
 - the number of problem1, problem2, and problem3 initial states are 12, 27 and 32. The larger the number of state is, The large the number of expansions is. As AIND 10.2.1 says, the total cost for naive algorithm estimates (average possible actions)**(path length). So, the cost order tends to be exponential for the number of cargo in goal states.
- breadth-first search(BFS)
 - BFS could find the optimal path for each problem. This algorithm can becomes a benchmark to compare to other algorithms because of the simple strategy.
- depth-first search(DFS)
 - DFS could not find optimal path. This is because this search algorithm finished when goal is found, regardless of the goal is optimal. And therefore, this algorithm is the fastest(minimum elapsed time) among I choose.
- uniform-cost search(UCS)
 - UCS could find the optimal path for each problem as lesson videos said. This algorithm's expansions and elapsed time were larger than breadth-first. However, the path cost of the UCS is same as the number of steps(uniform_cost_search's path_cost uses Problem.path_cost). So, essentially, UCS cost is same order as BFS cost in this implements.

heuristic search

-----	-----	-----	-----	-----	-----	-----	-----	-----	-----
-----	Problem	Algorithm	Path Length	Expansions	Elapsed Time (sec)		-----	-----	-----
-----	-----	-----	-----	-----	-----	-----	-----	-----	-----
----	Problem1	ignore preconditions	6 41 0.188	Problem1	level-sum	6 11	2.246		-----
-----	Problem2	ignore preconditions	9 1506 49.46	Problem2	level-sum	9 86 301		-----	-----
-----	Problem3	ignore preconditions	12 5118 421.81	Problem3	level-sum	12 404 2692		-----	-----

- ignore preconditions
 - This algorithm is the minimum elapsed time in my environment and expansions is lower than BFS. This algorithm is similar to A* search (estimate minimum distance ignoring wall). The decreased expansions designates the efficient path cut-off.
- level-sum
 - This algorithm is the minimum expansions. However the elapsed time is larger than BFS and ignore preconditions. I think this is because look-ahead search in heuristic function is higher cost. In other words, expansion is done in heuristic function. According to AIND 10.3.1, this algorithm is effective for largely decomposable problem. So this algorithm seems not be suitable for this problem (this problem's actions are largely constrained by each cargo and plain can be exists same places, not so decomposable).