

Planning Search Analysis

AIND-Planning Project

Non-heuristic Planning Solution Searches

For the air_cargo_p1, air_cargo_p2 and air_cargo_p3 problems, following algorithms are tested:

- breadth_first_search: Breadth-first search
- depth_first_graph_search: Depth-first search
- depth_limited_search: Depth-first search with the depth limitation of 50

Problem	Algorithms	Expansions	Goal Tests	Plan Length	Time Elapsed
Air Cargo P1	Breadth-first search	43	56	6	0.040s
	Depth-first search	21	22	20	0.016s
	Depth-first limited	101	271	50	0.109s
Air Cargo P2	Breadth-first search	3343	4609	9	15.284s
	Depth-first search	624	625	619	3.783s
	Depth-first limited	222719	2053741	50	1148.715s
Air Cargo P3	Breadth-first search	14663	18098	12	121.492s
	Depth-first search	408	409	392	2.031s
	Depth-first limited	NA	NA	NA	>20min

As shown in the above table, the metrics on number of node expansions required, number of goal tests, plan length of solution (breadth-first search always give the optimal plan length), and time elapsed for each search algorithm are shown.

Domain-independent Heuristics Searches

For the same three problems, following heuristics are tested with A* search:

- h_1: A constant value 1, not a real heuristic.
- h_ignore_preconditions: The minimum number of actions that must be carried out from the current state in order to satisfy all of the goal conditions by ignoring the preconditions required for an action to be executed.
- h_pg_levelsum: The sum of the level costs of the individual goals (admissible if goals independent). This heuristic can be inadmissible, but works well in real practice.

Problem	Heuristics	Expansions	Goal Tests	Plan Length	Time Elapsed
Air Cargo P1	constant 1	55	57	6	0.046s
	ignore precond	42	44	6	0.141s
	level sum	11	13	6	0.790s
Air Cargo P2	constant 1	4826	4828	9	15.025s
	ignore precond	2683	2685	9	89.280s
	level sum	86	88	9	79.304s
Air Cargo P3	constant 1	18221	18223	12	68.232s
	ignore precond	10244	10246	12	800.204s
	level sum	310	312	12	431.789s

As shown in the above table, the metrics on number of node expansions required, number of goal tests, plan length of solution (breadth-first search always give the optimal plan length), and time elapsed for each search algorithm are shown.

Optimal Plans

In this section we show the optimal results from A* search using h_pg_levelsum heuristic.

Air Cargo P1

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

Air Cargo P2

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

Air Cargo P3

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

Comparisons among Search Algorithms and Heuristics

Firstly, for the non-heuristic searches in the first section, we can see that only breadth-first search gives the optimal solution. However, when looking at the running speed of the algorithms, depth-first search is much faster — though not giving the optimal solution, it explores much less nodes. The worst one is the depth-limited depth-first search: it usually find solution just at the limit depth, and enumerate lots of nodes — one can consider this behavior as an approximation of expanding the search tree to the limit depth, which is a terrible practice.

Secondly, for the three heuristics used in A* search, we can first see that A* search algorithm alone (with the fake h_1 heuristic) has a very close performance as the breadth-first search algorithm — they both outputs optimal solutions, and both explores similar number of nodes. When considering the nodes explored (expanded), h_ignore_preconditions have a smaller number, and h_pg_levelsum has the smallest of all. However, the calculation cost of these two heuristics are much higher, thus the running time not always decreases when one utilize the heuristics. For example, in the air cargo problem 3, we can find that though h_ignore_preconditions explores less nodes than h_1, but its running time is way longer; Meanwhile, h_pg_levelsum explores two

magnitude less nodes than h_1 and $h_{\text{ignore_preconditions}}$, its running time is in the middle of those of h_1 and $h_{\text{ignore_preconditions}}$. Though $h_{\text{pg_levelsum}}$ is not admissible in our problem, but it works well since the goals are still highly decomposable.

Both the number of expanded nodes and the calculation cost on each node matter. Sometimes one should make tradeoff between these two: Better heuristics usually takes more time on each node, but it could help reduce the number of expansion; Vice versa, when expected to explore a large space, the algorithm cannot spend too much time on each node and have to make quick decisions. Roughly speaking, one can estimate the time by the product of expected node expansion number and the calculation complexity of heuristic function on each node. Furthermore, despite the time complexity, one should also take the space complexity in mind, especially when the memory usage is highly correlated with the size of expanded nodes. For example, in both breath-first and A* searches, the “frontier” of the searching nodes are kept in memory; In A* search, one may also want to cache some of the heuristic function results. Along this dimension of tradeoff, depth-first search makes much less burden on the space complexity.

So there is no one-for-all best combination of search algorithm and heuristic for problems. For finding the optimal solution in Air Cargo Problem, when the search space is quite small (e.g. Air Cargo Problem 1), simple breath-first search is enough and fast to find optimal solution. When the possible search space is quite large, A* search can perform better than breath-first search. Then, simple heuristics would lead to quick running time but larger memory usage, while complex heuristics like $h_{\text{pg_levelsum}}$ can dramatically reduce the number of expanded nodes, while compensate for the longer running time.