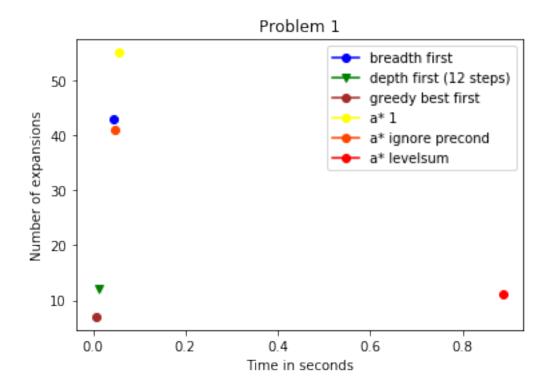
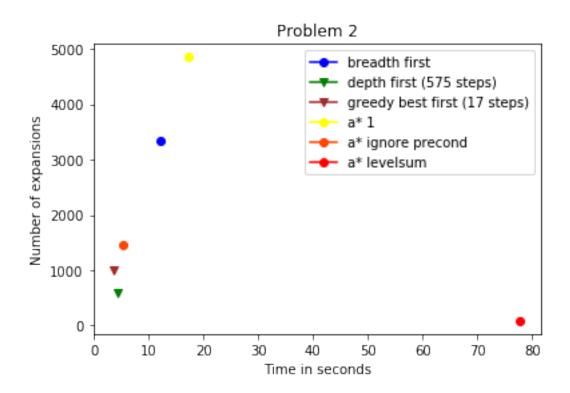
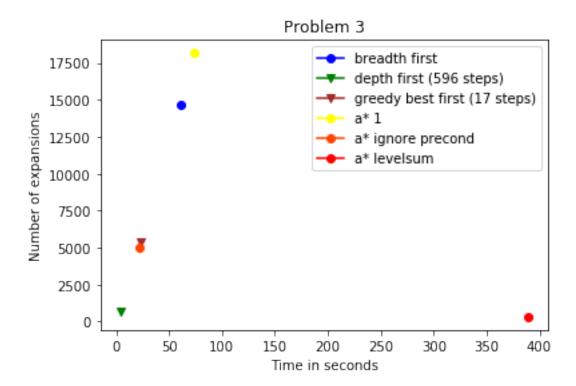
Comparison of search algorithms

Here is the result for a subset of the algorithms. Dots are used when the solution found is optimal. If the solution is not optimal, a triangle is used. The program output that was used to create these diagrams can be found in Appendix 2 at the end of the document.







The optimal solutions for problems 1, 2, and 3 contain 6, 9, and 12 steps, respectively. depth_first and greedy_best_first find solutions that are not optimal for problems 2 and 3.

Interpretation

Breadth first tree search

By definition, breadth first yields optimal solutions. There are many expansions because all states with path length < solution length have to be created. The speed is in the medium range. According to the AIMA book (Section 3.4.1), execution time can be a problem for breadth first. Here this is no problem because the search space is small.

Depth first graph search

Depth first goes all the way down the graph and backtracks only if no more nodes exist. Therefore the path length is almost as high as the number of expansions, and the solutions are far from optimal. As this is depth first *graph* search, the algorithm does not select the same state twice - it just keeps exploring new states until a solution is found. Therefore the algorithm is fast and needs relatively few expansions.

Greedy best first graph search with dummy heuristic

As the heuristic function returns the same value for every node, this algorithm selects nodes randomly. This works surprinsingly well: the algorithm is fast, and the solutions for problem 2 and 3 are not optimal, but much better than those from depth_first.

A* search with dummy heuristic

The A* heuristic estimates total path length as the sum of the path so far and the value of the heuristic function. As the heuristic here always returns 1, this algorithm is equivalent to uniform_cost (see section 3.5.2 in the AIMA book). This means that solutions are optimal. Speed is the second lowest of the algorithms considered, and the number of expansions is the highest. The reason that speed and number of expansions are worse than for breadth first is pointed out in section 3.4.2 of the AIMA book: as uniform cost guarantees to find an optimal solution, the goal test is applied to a node only before it is expanded, not directly after it has been created.

A* search with ignore preconditions heuristic

As pointed out in section 10.2.3 of the AIMA book, this heuristic does not guarantee to find an optimal solution. But in our case the solutions are optimal. The number of expansions is medium, and the speed is good. Given the data, this algorithm works best overall.

A* search with levelsum heuristic

As mentioned in section 10.3.1 of the AIMA book, this algorithm also does not guarantee optimal solutions. But like ignore_preconditions, optimal solutions are found for our problems. This algorithm is four times slower than the slowest of the others. It is so slow because the planning graph is constructed in each call of the heuristic function.

As the planning graph has polynomial size, this heuristic should work comparatively well for large search spaces. The problems considered here are so simple that the additional effort does not pay off.

Final thoughts

It is not surprising that using an informed heuristic like ignore_preconditions or levelsum yields better results than uninformed searches. But as a general note, I am surprised by the fact that automated planning is so difficult. The problems that we considered here look very simple – a human can solve them in a few seconds. The algorithms we looked at are not better than that. In contrast, the algorithms we implemented for the Sudoku project and the Isolation project outperformed humans.

Appendix 1: optimal plans

Problem 1

```
Load(C1, P1, SF0)
Load(C2, P2, JFK)
Fly(P1, SF0, JFK)
Fly(P2, JFK, SF0)
Unload(C1, P1, JFK)
Unload(C2, P2, SF0)
```

Problem 2

```
Load(C1, P1, SF0)
Load(C2, P2, JFK)
```

```
Load(C3, P3, ATL)
Fly(P1, SFO, JFK)
Fly(P2, JFK, SF0)
Fly(P3, ATL, SF0)
Unload(C1, P1, JFK)
Unload(C2, P2, SF0)
Unload(C3, P3, SF0)
Problem 3
Load(C2, P2, JFK)
Fly(P2, JFK, ORD)
Load(C4, P2, ORD)
Fly(P2, ORD, SF0)
Load(C1, P1, SF0)
Fly(P1, SFO, ATL)
Load(C3, P1, ATL)
Fly(P1, ATL, JFK)
Unload(C1, P1, JFK)
Unload(C3, P1, JFK)
Unload(C2, P2, SF0)
Unload(C4, P2, SF0)
Appendix 2: program output
Problem 1:
1) Solving Air Cargo Problem 1 using breadth_first_search...
Expansions
             Goal Tests
                          New Nodes
    43
                56
                           180
Plan length: 6 Time elapsed in seconds: 0.045388131432586076
Load(C2, P2, JFK)
Load(C1, P1, SF0)
Fly(P2, JFK, SF0)
Unload(C2, P2, SF0)
Fly(P1, SFO, JFK)
Unload(C1, P1, JFK)
3) Solving Air Cargo Problem 1 using depth_first_graph_search...
Expansions
             Goal Tests
                          New Nodes
    12
                13
                            48
Plan length: 12 Time elapsed in seconds: 0.011834418901481625
Fly(P1, SFO, JFK)
Fly(P2, JFK, SFO)
Load(C1, P2, SF0)
Fly(P2, SFO, JFK)
Fly(P1, JFK, SF0)
```

```
Unload(C1, P2, JFK)
Fly(P2, JFK, SFO)
Fly(P1, SFO, JFK)
Load(C2, P1, JFK)
Fly(P2, SFO, JFK)
Fly(P1, JFK, SFO)
Unload(C2, P1, SF0)
7) Solving Air Cargo Problem 1 using greedy_best_first_graph_search with h_1...
Expansions
             Goal Tests
                          New Nodes
    7
                9
                             28
Plan length: 6 Time elapsed in seconds: 0.007335399038430168
Load(C1, P1, SF0)
Load(C2, P2, JFK)
Fly(P1, SFO, JFK)
Fly(P2, JFK, SF0)
Unload(C1, P1, JFK)
Unload(C2, P2, SF0)
8) Solving Air Cargo Problem 1 using astar_search with h_1...
             Goal Tests
Expansions
                          New Nodes
    55
                57
                           224
Plan length: 6 Time elapsed in seconds: 0.05687500058313688
Load(C1, P1, SF0)
Load(C2, P2, JFK)
Fly(P1, SFO, JFK)
Fly(P2, JFK, SFO)
Unload(C1, P1, JFK)
Unload(C2, P2, SF0)
9) Solving Air Cargo Problem 1 using astar_search with h_ignore_preconditions...
Expansions
             Goal Tests
                          New Nodes
    41
                43
                           170
Plan length: 6 Time elapsed in seconds: 0.04603751298063331
Load(C1, P1, SF0)
Fly(P1, SFO, JFK)
Unload(C1, P1, JFK)
Load(C2, P2, JFK)
Fly(P2, JFK, SFO)
Unload(C2, P2, SF0)
10) Solving Air Cargo Problem 1 using astar_search with h_pg_levelsum...
Expansions
             Goal Tests
                          New Nodes
                             50
    11
                13
```

```
Plan length: 6 Time elapsed in seconds: 0.8866269386978214
Load(C1, P1, SF0)
Fly(P1, SFO, JFK)
Load(C2, P2, JFK)
Fly(P2, JFK, SF0)
Unload(C1, P1, JFK)
Unload(C2, P2, SF0)
Problem 2:
$ python run_search.py -p 2 -s 1
Solving Air Cargo Problem 2 using breadth_first_search...
             Goal Tests
                          New Nodes
Expansions
   3343
               4609
                          30509
Plan length: 9 Time elapsed in seconds: 12.090223915341547
Load(C2, P2, JFK)
Load(C1, P1, SF0)
Load(C3, P3, ATL)
Fly(P2, JFK, SFO)
Unload(C2, P2, SF0)
Fly(P1, SFO, JFK)
Unload(C1, P1, JFK)
Fly(P3, ATL, SF0)
Unload(C3, P3, SF0)
$ python run_search.py -p 2 -s 3
Solving Air Cargo Problem 2 using depth_first_graph_search...
                          New Nodes
Expansions
             Goal Tests
   582
               583
                           5211
Plan length: 575 Time elapsed in seconds: 4.3784089032449485
$ python run_search.py -p 2 -s 7
Solving Air Cargo Problem 2 using greedy_best_first_graph_search with h_1...
Expansions
             Goal Tests
                          New Nodes
   998
               1000
                           8986
Plan length: 17 Time elapsed in seconds: 3.576133921882012
Load(C1, P1, SF0)
Fly(P1, SFO, ATL)
Load(C2, P2, JFK)
Fly(P2, JFK, ATL)
```

```
Load(C3, P3, ATL)
Fly(P3, ATL, JFK)
Fly(P1, ATL, JFK)
Unload(C1, P1, JFK)
Load(C1, P3, JFK)
Fly(P1, JFK, ATL)
Fly(P2, ATL, SF0)
Unload(C2, P2, SF0)
Fly(P2, SF0, ATL)
Fly(P3, JFK, SFO)
Unload(C3, P3, SF0)
Fly(P3, SFO, JFK)
Unload(C1, P3, JFK)
$ python run_search.py -p 2 -s 8
Solving Air Cargo Problem 2 using astar_search with h_1...
Expansions
             Goal Tests
                          New Nodes
   4853
               4855
                          44041
Plan length: 9 Time elapsed in seconds: 17.222562463320674
Load(C1, P1, SF0)
Fly(P1, SFO, JFK)
Load(C2, P2, JFK)
Fly(P2, JFK, SF0)
Load(C3, P3, ATL)
Fly(P3, ATL, SF0)
Unload(C1, P1, JFK)
Unload(C3, P3, SF0)
Unload(C2, P2, SF0)
Solving Air Cargo Problem 2 using astar_search with h_ignore_preconditions...
Expansions
             Goal Tests
                          New Nodes
   1450
               1452
                          13303
Plan length: 9 Time elapsed in seconds: 5.394788526473028
Load(C1, P1, SF0)
Fly(P1, SFO, JFK)
Unload(C1, P1, JFK)
Load(C3, P3, ATL)
Fly(P3, ATL, SFO)
Unload(C3, P3, SF0)
Load(C2, P2, JFK)
Fly(P2, JFK, SF0)
Unload(C2, P2, SF0)
$ python run_search.py -p 2 -s 10
```

Solving Air Cargo Problem 2 using astar_search with h_pg_levelsum...

```
86
                          841
Plan length: 9 Time elapsed in seconds: 77.78207557644275
Load(C1, P1, SF0)
Fly(P1, SFO, JFK)
Load(C2, P2, JFK)
Fly(P2, JFK, SFO)
Load(C3, P3, ATL)
Fly(P3, ATL, SF0)
Unload(C1, P1, JFK)
Unload(C3, P3, SF0)
Unload(C2, P2, SF0)
-----
Problem 3:
$ python run_search.py -p 3 -s 1
Solving Air Cargo Problem 3 using breadth_first_search...
Expansions
            Goal Tests
                         New Nodes
  14663
              18098
                         129631
Plan length: 12 Time elapsed in seconds: 60.87180639099674
Load(C2, P2, JFK)
Load(C1, P1, SF0)
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, SF0)
Unload(C2, P2, SF0)
Unload(C4, P2, SF0)
$ python run_search.py -p 3 -s 3
Solving Air Cargo Problem 3 using depth_first_graph_search...
Expansions
            Goal Tests
                        New Nodes
                          5176
   627
               628
Plan length: 596 Time elapsed in seconds: 4.597075077277337
$ python run_search.py -p 3 -s 7
Solving Air Cargo Problem 3 using greedy_best_first_graph_search with h_1...
```

New Nodes

Goal Tests

Expansions

```
5399
               5401
                          47691
Plan length: 17 Time elapsed in seconds: 22.71502973066527
Load(C1, P1, SF0)
Fly(P1, SF0, ORD)
Load(C2, P2, JFK)
Fly(P2, JFK, ORD)
Load(C4, P2, ORD)
Fly(P2, ORD, ATL)
Load(C3, P2, ATL)
Fly(P2, ATL, ORD)
Unload(C3, P2, ORD)
Load(C3, P1, ORD)
Fly(P1, ORD, JFK)
Unload(C3, P1, JFK)
Unload(C1, P1, JFK)
Fly(P1, JFK, ORD)
Fly(P2, ORD, SF0)
Unload(C4, P2, SF0)
Unload(C2, P2, SF0)
$ python run_search.py -p 3 -s 8
Solving Air Cargo Problem 3 using astar_search with h_1...
Expansions
             Goal Tests
                          New Nodes
              18166
  18164
                          159147
Plan length: 12 Time elapsed in seconds: 73.68106735560823
Load(C1, P1, SF0)
Fly(P1, SFO, ATL)
Load(C2, P2, JFK)
Fly(P2, JFK, ORD)
Load(C3, P1, ATL)
Load(C4, P2, ORD)
Fly(P2, ORD, SF0)
Fly(P1, ATL, JFK)
Unload(C3, P1, JFK)
Unload(C1, P1, JFK)
Unload(C4, P2, SF0)
Unload(C2, P2, SF0)
Solving Air Cargo Problem 3 using astar_search with h_ignore_preconditions...
Expansions
             Goal Tests
                          New Nodes
   5038
               5040
                          44924
Plan length: 12 Time elapsed in seconds: 21.994492386094652
Load(C1, P1, SF0)
```

Goal Tests

New Nodes

Expansions

```
Fly(P1, SFO, ATL)
Load(C3, P1, ATL)
Fly(P1, ATL, JFK)
Unload(C3, P1, JFK)
Unload(C1, P1, JFK)
Load(C2, P2, JFK)
Fly(P2, JFK, ORD)
Load(C4, P2, ORD)
Fly(P2, ORD, SF0)
Unload(C4, P2, SF0)
Unload(C2, P2, SF0)
python run_search.py -p 3 -s 10
             Goal Tests
                          New Nodes
Expansions
   316
               318
                           2913
Plan length: 12 Time elapsed in seconds: 388.75686865605076
Load(C2, P2, JFK)
Fly(P2, JFK, ORD)
Load(C4, P2, ORD)
Fly(P2, ORD, SF0)
Load(C1, P1, SF0)
Fly(P1, SFO, ATL)
Load(C3, P1, ATL)
Fly(P1, ATL, JFK)
Unload(C3, P1, JFK)
Unload(C1, P1, JFK)
```

Unload(C4, P2, SF0)
Unload(C2, P2, SF0)