

# Search Performance Comparison

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An analysis of the performance of search algorithms solving cargo problems

This is part of the Udacity nanodegree for Artificial Intelligence. The complete assignment can be found in the [README](#) to this repository.

This analysis covers three different problems that are solved with a variety of algorithms.

## Problem Space

All three problems are in the same domain of cargo logistics at airports. The underlying **Action Schema** is the following

```
Action(Load(c, p, a),  
      PRECOND: At(c, a) ∧ At(p, a) ∧ Cargo(c) ∧  
      Plane(p) ∧ Airport(a)  
      EFFECT: ¬ At(c, a) ∧ In(c, p))  
Action(Unload(c, p, a),  
      PRECOND: In(c, p) ∧ At(p, a) ∧ Cargo(c) ∧  
      Plane(p) ∧ 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))
```

where the problems are defined as

### Problem 1

```
Init(At(C1, SF0)  $\wedge$  At(C2, JFK)
       $\wedge$  At(P1, SF0)  $\wedge$  At(P2, JFK)
       $\wedge$  Cargo(C1)  $\wedge$  Cargo(C2)
       $\wedge$  Plane(P1)  $\wedge$  Plane(P2)
       $\wedge$  Airport(JFK)  $\wedge$  Airport(SF0))
Goal(At(C1, JFK)  $\wedge$  At(C2, SF0))
```

### Problem 2

```
Init(At(C1, SF0)  $\wedge$  At(C2, JFK)  $\wedge$  At(C3, ATL)
       $\wedge$  At(P1, SF0)  $\wedge$  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(SF0)  $\wedge$  Airport(ATL))
Goal(At(C1, JFK)  $\wedge$  At(C2, SF0)  $\wedge$  At(C3, SF0))
```

### Problem 3

```
Init(At(C1, SF0)  $\wedge$  At(C2, JFK)  $\wedge$  At(C3, ATL)  $\wedge$  At(C4,
ORD)
       $\wedge$  At(P1, SF0)  $\wedge$  At(P2, JFK)
       $\wedge$  Cargo(C1)  $\wedge$  Cargo(C2)  $\wedge$  Cargo(C3)  $\wedge$ 
```

```

Cargo(C4)
    ∧ Plane(P1) ∧ Plane(P2)
    ∧ Airport(JFK) ∧ Airport(SFO) ∧ Airport(ATL)
    ∧ Airport(ORD))
Goal(At(C1, JFK) ∧ At(C3, JFK) ∧ At(C2, SFO) ∧ At(C4,
SFO))

```

## Used Algorithms

The following algorithms are used to solve the above problems.

1. [Breadth First Search \(BFS\)](#)
2. [Depth First Graph Search \(DFS\)](#)
3. [Depth Limited Search \(DLS\)](#)
4. [Uniform Cost Search \(UCS\)](#)
5. [Greedy Best First Search \(GBFS\)](#)

## Comparison

The complete combination of problems and algorithms can be run with  
[python run\\_search.py -p 1 2 3 -s 1 3 4 5 7](#)

Problem	Algorithm	Expansions	Goal Tests	Nodes	Plan Length
1	BFS	43	56	180	6
1	DFS	12	13	48	12
1	DLS	103	271	414	50
1	UCS	55	57	224	6
1	GBFS	7	9	28	6
2	BFS	3,343	4,609	30,509	9
2	DFS	582	583	5211	575

2	DLS	222,719	2,053,741	205,4119	50
2	UCS	4,852	4,854	44,030	9
2	GBFS	990	992	8,910	15
3	BFS	14,663	18,098	129,631	12
3	DFS	627	628	5,176	596
3	DLS	–	–	–	–
3	UCS	18,164	18,166	1591,47	12
3	GBFS	5,399	5,401	47,691	17

## Observations

One thing that clearly shows in the comparison table is that algorithms that only run a small number of expansions and therefore cover smaller number of nodes lead to longer paths. This is expected because without a proper heuristic the algorithm would need to cover as much solution space as possible to find the best answer.

Those algorithms that yield the best results have a high number of iterated nodes. Breadth First and Uniform Cost Search seem to perform benchmark when looking at the path length. Depth First seems to deliver the fastest solution but in complex scenarios like Cargo 2 & 3 it delivers a really costly solution. Greedy Best First seems to provide a good tradeoff between solution speed and the cost of the calculated solution.

To my surprise the Depth Limited Search performs really bad in both categories. It takes longer than other algorithms to find a solution and the solution has one of the longest paths.

## Conclusion

With the observations made from this comparison the following

conclusions can be made:

1. For scenarios where the search duration needs to be minimized the **Greedy Best First** Algorithm delivers solutions quickly, however not the best solution
2. If the application allows for more time during the search the algorithms **Breadth First** or **Uniform Cost Search** deliver the solution with the lowest cost (shortest path).