## Heuristic Analysis

# PLANNING SEARCH

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Synopsis

In this project, we implemented a planning search agent to solve deterministic planning problem for an Air Cargo Transport system.

We use **planning graph** and **automatic domain independent heuristic with A\* search** and compare their results/performance against several **uninformed heuristic search methods.**

Planning Problem:

We were given 3 planning problems in Air Cargo domain with Action schema:

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) ∧ ¬ In (c, p))

Action (Fly(p, from, to),

PRECOND: At (p, from) ∧ Plane(p) ∧ Airport(from) ∧ Airport(to)

EFFECT: ¬ At (p, from) ∧ At(p, to))

Initial state and Goal and these 3 problems were:

Problem 1:

Init (At(C1, SFO) ∧ At(C2, JFK)

∧ At(P1, SFO) ∧ At(P2, JFK)

∧ Cargo(C1) ∧ Cargo(C2)

∧ Plane(P1) ∧ Plane(P2)

∧ Airport(JFK) ∧ Airport(SFO))

Goal (At(C1, JFK) ∧ At(C2, SFO))

Problem 2:

Init (At(C1, SFO) ∧ At(C2, JFK) ∧ At(C3, ATL)

∧ At(P1, SFO) ∧ At(P2, JFK) ∧ At(P3, ATL)

∧ Cargo(C1) ∧ Cargo(C2) ∧ Cargo(C3)

∧ Plane(P1) ∧ Plane(P2) ∧ Plane(P3)

∧ Airport(JFK) ∧ Airport(SFO) ∧ Airport(ATL))

Goal (At(C1, JFK) ∧ At(C2, SFO) ∧ At(C3, SFO))

Problem 3:

Init (At(C1, SFO) ∧ At(C2, JFK) ∧ At(C3, ATL) ∧ At(C4, ORD)

∧ At(P1, SFO) ∧ At(P2, JFK)

∧ Cargo(C1) ∧ Cargo(C2) ∧ Cargo(C3) ∧ 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))

The goals described above can be reached using different plans.

Optimal path length of problem 1, 2 and 3 are 6, 9 and 12 respectively. Below are the actions taken in each problem to reach goal state:

**Problem 1:**

Load(C1, P1, SFO)

Load(C2, P2, JFK)

Fly(P1, SFO, JFK)

Fly(P2, JFK, SFO)

Unload(C1, P1, JFK)

Unload(C2, P2, SFO)

**Problem 2:**

Load(C1, P1, SFO)

Load(C2, P2, JFK)

Load(C3, P3, ATL)

Fly(P1, SFO, JFK)

Fly(P2, JFK, SFO)

Fly(P3, ATL, SFO)

Unload(C3, P3, SFO)

Unload(C2, P2, SFO)

Unload(C1, P1, JFK)

Problem 3:

Load(C1, P1, SFO)

Load(C2, P2, JFK)

Fly(P1, SFO, ATL)

Load(C3, P1, ATL)

Fly(P2, JFK, ORD)

Load(C4, P2, ORD)

Fly(P2, ORD, SFO)

Fly(P1, ATL, JFK)

Unload(C4, P2, SFO)

Unload(C3, P1, JFK)

Unload(C2, P2, SFO)

Unload(C1, P1, JFK)

**Uninformed search strategies:**

The term uniformed search strategies means that the strategies have no additional information about the states beyond that provided in problem definition. All they can do is generate successors and distinguish a goal state from non-goal state. All search strategies are distinguished by the order in which nodes are expanded.

In this section, we compare the performance of such strategies in terms of speed, memory usage and optimality.

Below are the performance results for the same:

Problem 1 result:

|  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- |
| **Search Type** | **Optimal** | **Path Length** | **Execution time** | **Node expansion** | **Goal Test** | **New nodes** |
| Breadth first search | Yes | 6 | 0.08 | 43 | 56 | 180 |
| Breadth first tree search | Yes | 6 | 1.10 | 1458 | 1459 | 5960 |
| Depth first graph search | No | 12 | 0.008 | 12 | 13 | 48 |
| Depth limited search | No | 50 | 0.098 | 101 | 271 | 414 |
| Uniform cost search | Yes | 6 | 0.040 | 55 | 57 | 224 |
| Recursive best first search | Yes | 6 | 3.08 | 4229 | 4230 | 17029 |
| **Greedy Best First GS with h\_1** | **Yes** | **6** | **0.005** | **7** | **9** | **28** |

Problem 2 result:

In problem 2 we cancelled data collection for Breadth first tree search; depth limited search and recursive best first search because their execution time was exceeding 10 min.

|  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- |
| **Search Type** | **Optimal** | **Path Length** | **Execution time** | **Node expansion** | **Goal Test** | **New nodes** |
| Breadth first search | **Yes** | **9** | 15.41 | 3346 | 4612 | 30534 |
| Breadth first tree search | - | - | - | - | - | - |
| Depth first graph search | No | 1085 | **9.59** | **1124** | **1125** | **10017** |
| Depth limited search | - | - | - | - | - | - |
| Uniform cost search | **Yes** | **9** | 29.03 | 4852 | 4854 | 44030 |
| Recursive best first search | - | - | - | - | - | - |
| Greedy Best First GS with h\_1 | Yes | 21 | **3.11** | **990** | **992** | **8910** |

Problem 3 results:

In problem 3 we cancelled the data collection for breadth first tree search, Depth limited search and recursive best first search because their execution time was exceeding 10 min.

|  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- |
| **Search Type** | **Optimal** | **Path Length** | **Execution time** | **Node expansion** | **Goal Test** | **New nodes** |
| Breadth first search | **Yes** | **12** | 125.76 | 14120 | 17673 | 123964 |
| Breadth first tree search | - | - | - | - | - | - |
| Depth first graph search | No | 2031 | **46.31** | **5591** | **5592** | **45563** |
| Depth limited search | - | - | - | - | - | - |
| Uniform cost search | **Yes** | **12** | 74.13 | 18235 | 18237 | 158272 |
| Recursive best first search | - | - | - | - | - | - |
| Greedy Best First GS with h\_1 | Yes | 26 | **20.17** | **5673** | **5675** | **49221** |

Analysis:

With the 3-problem set we found that **Breadth first search** **and Uniform cost search** are the only two uninformed strategies that provided optimal action plan in terms of all the 3 parameters speed (exc. Time), memory usage(no of nodes expended) and optimality (find goal state). **Depth first search** is good in terms of execution time, node expansion but not good in terms of optimality, as we can see in all the problem sets path length derived are 12, 1085 and 2031 which is not good in comparison to 6, 9 and 12 respectively. So if the main motive is speed and memory Depth First search can be given priority.

In case of finding **optimal** **path** **Breadth first search** should be given priority because it performs **faster** and **uses less memory** than uniform cost search.

**Informed search strategies analysis:**

Informed search strategy is the one that uses problem specific knowledge beyond the definition of problem itself – can find solutions more efficiently than uniformed strategies.

Below are the performance measures collected for problem 1, 2 and 3 using informed search strategies.

Problem 1 Results:

|  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- |
| **Search Type** | **Optimal** | **Path Length** | **Execution time** | **Node expansion** | **Goal Test** | **New nodes** |
| A\* with h\_1 | Yes | 6 | 0.03 | 55 | 57 | 224 |
| A\* with h\_ignore\_prec. | Yes | **6** | **0.044** | 41 | 43 | 170 |
| A\* with h\_pg\_levelsum | Yes | 6 | 0.9 | **11** | **13** | **50** |

Problem 2 Results:

|  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- |
| **Search Type** | **Optimal** | **Path Length** | **Execution time** | **Node expansion** | **Goal Test** | **New nodes** |
| A\* with h\_1 | Yes | 9 | 15.06 | 4852 | 4854 | 44030 |
| A\* with h\_ignore\_prec. | Yes | **9** | **5.34** | 1450 | 1452 | 13303 |
| A\* with h\_pg\_levelsum | Yes | 9 | 201.39 | **86** | **88** | **841** |

Problem 3 Results:

|  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- |
| **Search Type** | **Optimal** | **Path Length** | **Execution time** | **Node expansion** | **Goal Test** | **New nodes** |
| A\* with h\_1 | Yes | 12 | 65.47 | 18235 | 18237 | 158272 |
| A\* with h\_ignore\_prec. | Yes | **12** | **23.32** | 5040 | 5042 | 44769 |
| A\* with h\_pg\_levelsum | Yes | 12 | 1303.53 | **387** | **389** | **3550** |

Analysis:

While all heuristics yield an optimal action plan, only h1 and ignore precondition heuristics return results within 10 min max execution time. If we consider speed and optimal path length then **A\* with ignore precondition heuristics** should be used as we can see in above data collection although A\* with heuristics and A\* with ignore precondition heuristics yield result with same path length but execution time of A\* with ignore heuristics is much more efficient in execution time.

If we consider memory than A\* search with level sum heuristics uses least memory but execution time much more as compared to others.

Comparison of informed and uninformed search strategies:

Problem 1:

|  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- |
| **Search Type** | **Optimal** | **Path Length** | **Execution time** | **Node expansion** | **Goal Test** | **New nodes** |
| Breadth first search | Yes | 6 | 0.08 | 43 | 56 | 180 |
| **A\* with h\_ignore\_prec.** | **Yes** | **6** | **0.044** | **41** | **43** | **170** |

Problem 2:

|  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- |
| **Search Type** | **Optimal** | **Path Length** | **Execution time** | **Node expansion** | **Goal Test** | **New nodes** |
| Breadth first search | Yes | 9 | 15.41 | 3346 | 4612 | 30534 |
| **A\* with h\_ignore\_prec.** | **Yes** | **9** | **5.34** | **1450** | **1452** | **13303** |

Problem 3:

|  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- |
| **Search Type** | **Optimal** | **Path Length** | **Execution time** | **Node expansion** | **Goal Test** | **New nodes** |
| Breadth first search | Yes | 12 | 125.76 | 14120 | 17673 | 123964 |
| **A\* with h\_ignore\_prec.** | **Yes** | **12** | **23.32** | **5040** | **5042** | **44769** |

Conclusion:

Overall if we compare uninformed and informed search strategies, we could see that informed search strategies with heuristic search is more efficient in terms of speed, optimality and memory.

**A\* search with ignore precondition heuristics is the better choice overall.**

Reference:

Artificial intelligence [Third edition] by Stuart Russell and Peter Norvig.