

# Heuristic Analysis - Planning Search

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In this project, we implement a planning search problem for Air Cargo System. There are two analysis results in this document: **non-heuristic planning** and **heuristic planning**.

## Planning problem

Here are all the action schema:

Action(Load(c, p, a),

PRECOND:  $At(c, a) \wedge At(p, a) \wedge Cargo(c) \wedge Plane(p) \wedge Airport(a)$

EFFECT:  $\neg At(c, a) \wedge In(c, p)$

Action(Unload(c, p, a),

PRECOND:  $In(c, p) \wedge At(p, a) \wedge Cargo(c) \wedge Plane(p) \wedge Airport(a)$

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:  $\neg At(p, from) \wedge At(p, to)$

There are three problems need to be solved:

- Problem 1

Init( $At(C1, SFO) \wedge At(C2, JFK)$

$\wedge At(P1, SFO) \wedge At(P2, JFK)$

$\wedge Cargo(C1) \wedge Cargo(C2)$

$\wedge Plane(P1) \wedge Plane(P2)$

$\wedge Airport(JFK) \wedge Airport(SFO)$ )

Goal( $At(C1, JFK) \wedge At(C2, SFO)$ )

- Problem 2

Init( $At(C1, SFO) \wedge At(C2, JFK) \wedge At(C3, ATL)$

$\wedge At(P1, SFO) \wedge At(P2, JFK) \wedge At(P3, ATL)$

$\wedge Cargo(C1) \wedge Cargo(C2) \wedge Cargo(C3)$

$\wedge Plane(P1) \wedge Plane(P2) \wedge Plane(P3)$

$\wedge Airport(JFK) \wedge Airport(SFO) \wedge Airport(ATL)$ )

Goal( $At(C1, JFK) \wedge At(C2, SFO) \wedge At(C3, SFO)$ )

- Problem 3

Init( $\text{At}(\text{C1}, \text{SFO}) \wedge \text{At}(\text{C2}, \text{JFK}) \wedge \text{At}(\text{C3}, \text{ATL}) \wedge \text{At}(\text{C4}, \text{ORD})$ )

$\wedge \text{At}(\text{P1}, \text{SFO}) \wedge \text{At}(\text{P2}, \text{JFK})$

$\wedge \text{Cargo}(\text{C1}) \wedge \text{Cargo}(\text{C2}) \wedge \text{Cargo}(\text{C3}) \wedge \text{Cargo}(\text{C4})$

$\wedge \text{Plane}(\text{P1}) \wedge \text{Plane}(\text{P2})$

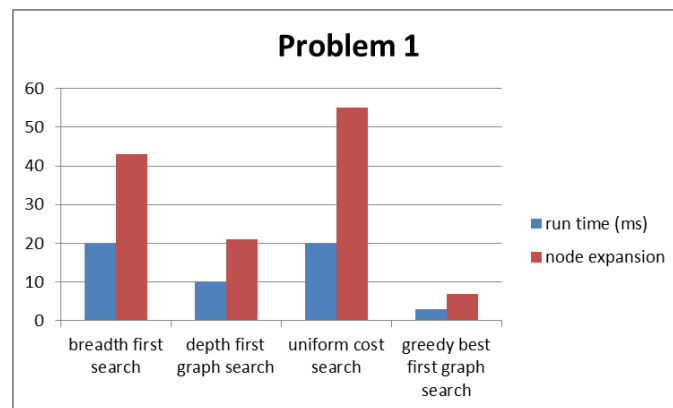
$\wedge \text{Airport}(\text{JFK}) \wedge \text{Airport}(\text{SFO}) \wedge \text{Airport}(\text{ATL}) \wedge \text{Airport}(\text{ORD})$ )

Goal( $\text{At}(\text{C1}, \text{JFK}) \wedge \text{At}(\text{C3}, \text{JFK}) \wedge \text{At}(\text{C2}, \text{SFO}) \wedge \text{At}(\text{C4}, \text{SFO})$ )

## non-heuristic planning result

- *problem 1*

search strategy	best plan length	run time (sec)	node expansion	result
breadth first search	6	0.020	43	
breadth first tree search	6	0.750	1458	
depth first graph search	20	0.010	21	x
depth limited search	50	0.070	101	x
uniform cost search	6	0.020	55	
recursively best first search	6	2.150	4226	
greedy best first graph search	6	0.003	7	best time & memory



In problem1, each search strategy tries to find its best plan. Node expansion means the number of nodes created. Therefore, it represents memory cost. In this result, **depth first graph search** and **depth limited search** didn't find the best plan. They are not good strategies even though they spend less time, because finding the best plan is more important. We think spend more time to get the best plan is acceptable. In this table, **greedy best first graph search** is the best strategy because it cost less time and memory usage.

## Optimal Plan

Load(C1, P1, SFO)

Load(C2, P2, JFK)

Load(C3, P3, ATL)

Fly(P2, JFK, SFO)

Unload(C2, P2, SFO)

Fly(P1, SFO, JFK)

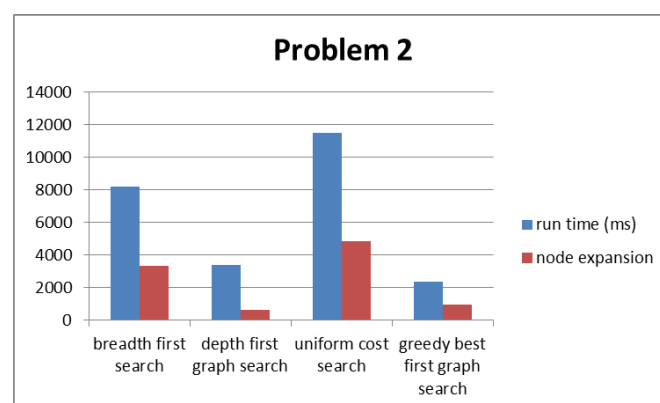
Unload(C1, P1, JFK)

Fly(P3, ATL, SFO)

Unload(C3, P3, SFO)

## ● problem 2

search strategy	best plan length	run time (sec)	node expansion	result
breadth first search	9	8.23	3343	best time and mem
breadth first tree search	-	-	-	
depth first graph search	619	3.39	624	x
depth limited search	-	-	-	
uniform cost search	9	11.50	4849	
recursively best first search	-	-	-	
greedy best first graph search	16	2.38	966	x



In problem 2, this is a more complex problem than problem 1. We disable the strategies which cost more than 10 minutes. Unfortunately, **greedy best first graph search** didn't found the best plan. **Breadth first search** found the best plan and cost less time and memory usage.

## Optimal Plan

Load(C1, P1, SFO)

Load(C2, P2, JFK)

Load(C3, P3, ATL)

Fly(P2, JFK, SFO)

Unload(C2, P2, SFO)

Fly(P1, SFO, JFK)

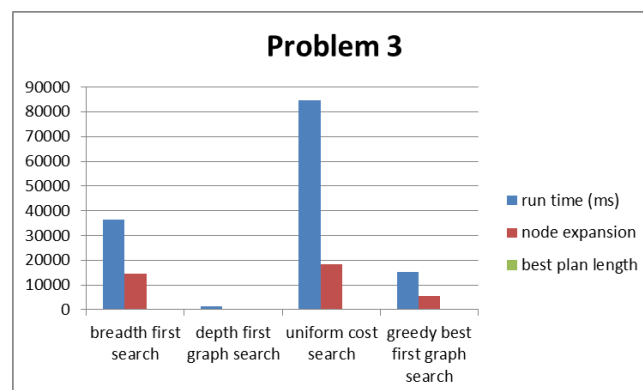
Unload(C1, P1, JFK)

Fly(P3, ATL, SFO)

Unload(C3, P3, SFO)

## ● problem 3

search strategy	best plan length	run time (sec)	node expansion	result
breadth first search	12	36.44	14663	best time and mem
breadth first tree search	-	-	-	
depth first graph search	392	1.52	408	x
depth limited search	-	-	-	
uniform cost search	12	84.80	18235	
recursively best first search	-	-	-	
greedy best first graph search	21	15.19	5462	x



Problem 3 is the most complex. Also, we disable the strategies which cost more than 10 minutes. **breadth first search** found the best plan and cost less time and memory usage.

### **Optimal Plan**

Load(C1, P1, SFO)

Load(C2, P2, JFK)

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)

## heuristic planning result

Heuristic strategy can find solutions more efficiently than non-heuristic strategy. We compare A\* search and add some different heuristics.

- *problem 1*

search strategy	best plan length	run time (sec)	node expansion	result
A* search	6	0.03	55	
A* search with ignore precondition	6	0.02	41	best time
A* search with level sum heuristic	6	0.59	11	best memory

- *problem 2*

search strategy	best plan length	run time (sec)	node expansion	result
A* search	9	9.99	4849	
A* search with ignore precondition	9	3.08	1443	best time
A* search with level sum heuristic	9	56.81	85	best memory

- *problem 3*

search strategy	best plan length	run time (sec)	node expansion	result
A* search	12	46.64	18235	
A* search with ignore precondition	12	14.03	4945	best time
A* search with level sum heuristic	12	249.44	290	best memory

**A\* search with ignore precondition** estimates the minimum number of actions that must be carried out from the current state in order to satisfy all of the goal conditions. Therefore, it costs less time than original A\* search.

**A\* search with level sum heuristic** uses a planning graph representation of the problem state space to estimate the sum of all actions that must be carried out from the current state in order to satisfy each individual goal condition. It costs more time because it needs to create the planning graph. When the problem is larger, the graph grows exponentially so that the run time increases tremendously. After building the graph, it can use level sum to reach the goal efficiently, therefore, the number node expansion become lower.

## Conclusion

search strategy	best plan length	run time (sec)	node expansion	result
breadth first search	12	36.44	14663	
A* search	12	46.64	18235	
A* search with ignore precondition	12	14.03	4945	best strategy
A* search with level sum heuristic	12	249.44	290	

In the book *Artificial Intelligence: A Modern Approach*, 3/e Chapter 10, "An admissible heuristic can be derived by defining a relaxed problem that is easier to solve. The exact cost of a solution to this easier problem then becomes the heuristic for the original problem". We compare heuristic and non-heuristic strategy in problem 3. In this table, heuristic method **A\* search with ignore precondition** has correct direction toward the goal rather than **breadth first search**. Therefore, it costs less time and memory space to reach the goal. When the problem become large, **A\* search with ignore precondition** has better performance than other methods.