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

## **Description Project:**

In this project, we implemented a planning search agent to solve deterministic logistics planning problems for an Air Cargo Transport System.

#### **Problems:**

The project contained 3 problems, this reporter is divided the analysis for each problem. The problems are:

```
\Lambda At (P1, SFO) \Lambda At (P2, JFK)
   Λ Cargo(C1) Λ Cargo(C2)
   \Lambda Plane(P1) \Lambda Plane(P2)
   Λ Airport(JFK) Λ Airport(SFO))
   Goal(At(C1, JFK) \land At(C2, SFO))
• Init(At(C1, SFO) / At(C2, JFK) / At(C3, ATL)
   \Lambda At (P1, SFO) \Lambda At (P2, JFK) \Lambda At (P3, ATL)
   Λ Cargo(C1) Λ Cargo(C2) Λ Cargo(C3)
   \Lambda Plane(P1) \Lambda Plane(P2) \Lambda Plane(P3)
   Λ Airport(JFK) Λ Airport(SFO) Λ Airport(ATL))
   Goal (At (C1, JFK) / At (C2, SFO) / At (C3, SFO))

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

   \Lambda At (P1, SFO) \Lambda At (P2, JFK)
   Λ Cargo(C1) Λ Cargo(C2) Λ Cargo(C3) Λ Cargo(C4)
   \Lambda Plane(P1) \Lambda Plane(P2)
   Λ Airport(JFK) Λ Airport(SFO) Λ Airport(ATL) Λ Airport(ORD))
```

## **Optimal Actions:**

The goal in every problem can be reached using different plans, the tables below show optimal planning for every problem.

Goal (At (C1, JFK)  $\Lambda$  At (C3, JFK)  $\Lambda$  At (C2, SFO)  $\Lambda$  At (C4, SFO))

Solution Problem 1

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

Solution Problem 2

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

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

### Solution Problem 3

Load(C1,P1,SFO)
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,SFO)
Unload(C4,P2,SFO)
Unload(C2,P2,SFO)

The previous tables show the optimal plan lengths for the **problem 1, 2 and 3 are 6, 9, 12 actions**; respectively.

## **Uninformed and Informed Search**

The following tables contain the results of experiments with different strategy, for the problem 1 used all strategies. But for the problem 2 and 3, in **Uninformed** only I used *breadth\_first\_search*, *depth\_first\_graph\_search*, *uniform\_cost\_search* and *greedy\_best\_first\_graph\_search* with h\_1.

In the tables I considered compare performance of the strategies in terms of speed (time elapsed in seconds), memory (Node Expansions and Goal Test) and Optimality (plan length).

#### Problem 1

Uninformed Strategy	Time Elapsed(s)	Plan Lenght	Node Expansions	Goal Test
breadth_first_search	0.057	6	43	56
breadth_first_tree_search	1.935	6	1458	1459
depth_first_graph_search	0.016	12	12	13
depth_limited_search	0.175	50	101	271
uniform_cost_search	0.066	6	55	57
recursive_best_first_search with h_1	5.341	6	4229	4230
greedy_best_first_graph_search with h_1	0.066	6	55	57

Informed Strategy	Time Elapsed (s)	Plan Lenght	Node Expansions	Goal Test
astar_search with h_1	0.070	6	55	57
astar_search with h_ignore_preconditions	0.079	6	41	43
astar_search with h_pg_levelsum	2.126	6	11	13

For the problem 1 we observed the all uninformed and informed strategies are optimal; plan length equal 6, except depth\_first\_graph\_search and depth\_limited\_search.

### Problem 2

Uninformed Strategy	Time Elapsed(s)	Plan Length	Node Expansions	Goal Test
breadth_first_search	27.205	9	3343	4609
depth_first_graph_search	7.654	575	582	583
uniform_cost_search	23.364	9	4761	4763
greedy_best_first_graph_search with h_1	3.004	9	550	552

Informed Strategy	Time Elapsed (s)	Plan Lenght	Node Expansions	Goal Test
astar_search with h_1	16.970	9	4761	4763
astar_search with h_ignore_preconditions	6.339	9	1450	1450
astar_search with h_pg_levelsum	286.476	9	86	88

For the problem2 we observed the all uninformed and informed strategies used are optimal: plan length equal 9, except *depth\_first\_graph\_search*.

## Problem 3

Uninformed Strategy	Time	Plan	Node	Goal
	Elapsed(s)	Length	Expansions	Test
breadth_first_search	198.658	12	14491	17947
depth_first_graph_search	39.711	1878	1948	1949
uniform_cost_search	90.973	12	17783	17785
greedy_best_first_graph_search with	23.291	22	4031	4033
h_1				

Informed Strategy	Time Elapsed (s)	Plan Length	Node Expansions	Goal Test
astar_search with h_1	77.682	12	17783	17785

astar_search with	25.887	12	5003	5005
h_ignore_preconditions				
astar_search with	1513.546	12	311	313
h_pg_levelsum				

For the problem 3 we observed these all uninformed and informed strategies used are optimal; plan Length equal 12, except *depth\_first\_graph\_search* and *greedy\_best\_first\_graph\_search* with *h\_1*.

## **Analysis of Results**

For the analysis only I considered the results in the table for the uninformed strategies: breadth\_first\_search, depth\_first\_graph\_search, uniform\_cost\_search and greedy\_best\_first\_graph\_search with h\_1. For the informed strategies, I used all strategies with heuristics.

For the graphics, I used the nomenclature for each strategy:

- Breadth first search =bfs
- Depth first graph search = dfgs
- Uniform cost search = ucs
- Greedy best first graph search=gbf
- Search with h\_1 = h\_1
- search with h\_ignore\_preconditions= h\_ignore
- search with h\_pg\_levelsum= h\_pg

### For Execution Time





The graphs have the scale in logarithms for "times elapsed". The first graph shows the strategies vs problems, it shows different behavior in problems. In the seconds graph show *log(time)* for every strategy in every problem.

In uninformed strategies show in the problem 2 and 3, the order in strategies is:

#### bfg>ucs>dfgs>gbf

Thus, **gbf** is the fastest planning search among bfg, ucs, dfgs and gbf.

In informed strategies show in the problem 2 and 3, the order in strategies is:

### h\_pg>h\_1>h\_ignore

Thus, **h\_ignore** is the fastest planning search. Comparing the two types of strategies; uninformed and informed, the **gbf** required less time in h\_ignore in the 3 problems.

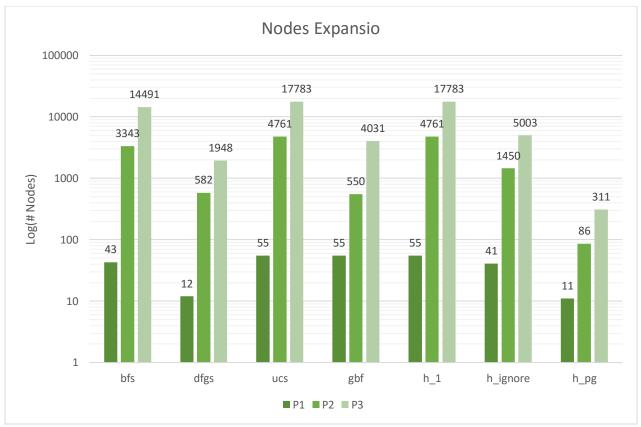
### For Node Expansion

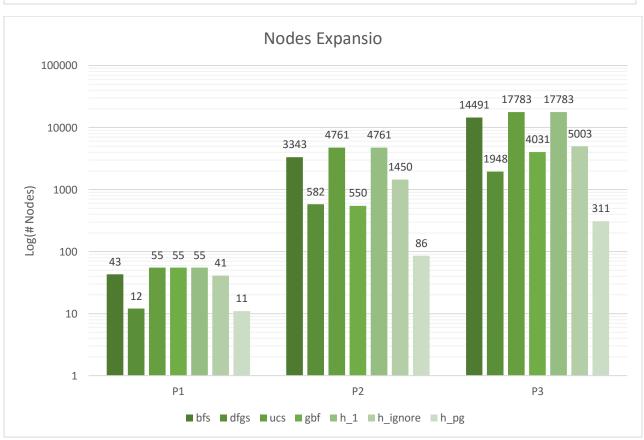
The first graph show node expansions strategies vs problems, it shows different behavior in problems. No is clear the relation between uninformed strategies vs problems, but **dfgs** and **gbf** require fewer nodes expansion.

In informed strategies show in the all problems, the order in strategies is:

### h\_1>h\_ignore>h\_pg

Thus, the number of nodes expanded is least by **h\_pg** heuristics.

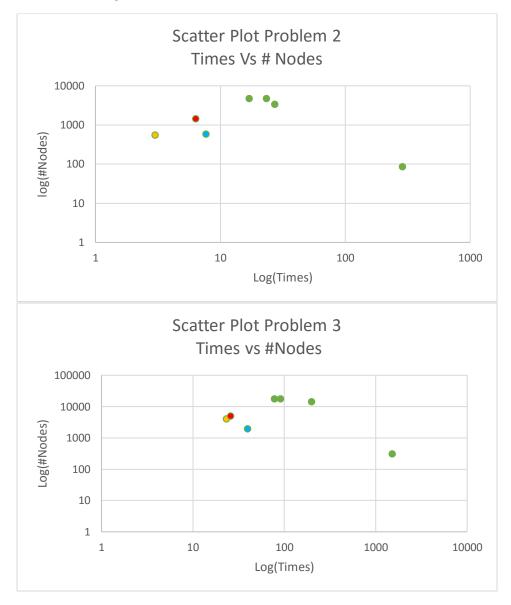




#### Conclusion

The optimality of each of the heuristics is guaranteed as all of them are admissible. All the heuristics under consideration provide optimal solution for each of the problems.

Comparing the strategies between the balance of #nodes vs times for problems 2 and 3 for each strategy we see the following:



The isolated point of the graph corresponds to the strategy **h\_pg**, it is appreciated that this have a low number of nodes but it requires a lot of time.

The following 3 strategies show a low number of nodes and a smaller amount of time, so a better balance is in red, blue and yellow. These correspond to the strategies h\_ingnore, dfgs and gbf, respectively. Of those 3 strategies, the only one that is optimal in all problems is h\_ignore, therefore it is the best strategy of both the uninformed and the informed.

# **Optimal Solution the A\*-Search using h\_ignore\_preconditions:**

### Problem 1

Load(C1, P1, SFO)

Fly(P1, SFO, JFK)

Unload(C1, P1, JFK)

Load(C2, P2, JFK)

Fly(P2, JFK, SFO)

Unload(C2, P2, SFO)

### Problem 2

Load(C3, P3, ATL)

Fly(P3, ATL, SFO)

Unload(C3, P3, SFO)

Load(C2, P2, JFK)

Fly(P2, JFK, SFO)

Unload(C2, P2, SFO)

Load(C1, P1, SFO)

Fly(P1, SFO, JFK)

Unload(C1, P1, JFK)

### **Problem 3**

Load(C2, P2, JFK)

Fly(P2, JFK, ORD)

Load(C4, P2, ORD)

Fly(P2, ORD, SFO)

Unload(C4, P2, SFO)

Load(C1, P1, SFO)

Fly(P1, SFO, ATL)

Load(C3, P1, ATL)

Fly(P1, ATL, JFK)

Unload(C3, P1, JFK)

Unload(C2, P2, SFO)

Unload(C1, P1, JFK)