

## ANALYSIS

### Description of the strategy:

Strategy 0: This strategy has no heuristic and choosing the next move randomly from the valid ones. It searches deeply into the random move at each step. The chance to get the right solution from these sequence of random picks is very small. Moreover, considering the high dimensionality, the machine buried itself deeply into the “wrong” and can rarely return a right solution.

Strategy 1: This strategy uses the simple heuristic based on the notion of fixed degree of the squares, which describes the number of neighbors on the board of each square. This strategy always pick randomly from the neighbors with the lowest fixed degree firstly to search for possible solutions.

Strategy 2: This strategy uses the simple heuristic based on the notion of dynamics degree of the squares, which describes the number of non-visited neighbors on the board of each square. This strategy always pick randomly from the neighbors with the lowest dynamics degree firstly to search for possible solutions.

Strategy 3: This strategy uses the simple heuristic based on the notion of fixed degree of the squares, which describes the number of neighbors on the board of each square. Different from strategy 1, before it moves to the next step, it will firstly check the dynamics degrees of the unvisited neighbors of the current square. If there are more than one such neighbor with the dynamics degree of 1, it will directly recognize it as a failure and return to its previous backup. Otherwise, this strategy always pick randomly from the neighbors with the lowest fixed degree firstly to search for possible solutions.

Strategy 4: This strategy uses the simple heuristic based on the notion of fixed degree of the squares, which describes the number of neighbors on the board of each square. Different from strategy 1 and 3, before it moves to the next step, it will firstly check the dynamics degrees of the unvisited neighbors of the current square. If there is exactly one such neighbor with the dynamics degree of 1, it will directly move into this square it and will not try any other neighbors after it returns from searching this neighbor (with dynamics degree of 1). Otherwise, this strategy always pick randomly from the neighbors with the lowest fixed degree firstly to search for possible solutions.

Strategy 5: This strategy is a combined version of 2,3,4.

This strategy uses the simple heuristic based on the notion of dynamics degree of the squares, which describes the number of neighbors on the board of each square. Different from strategy 1 and 3, before it moves to the next step, it will firstly check the dynamics degrees of the unvisited neighbors of the current square. If there are more than one such neighbor with the dynamics degree of 1, it will directly recognize it as a failure and return to its previous backup. If there is exactly one such neighbor with the dynamics degree of 1, it will directly move into this square it and will not try any other neighbors after it returns from searching this neighbor (with dynamics degree of 1). Otherwise, this strategy always pick randomly from the neighbors with the lowest dynamics degree firstly to search for possible solutions.

Strategy 6: (My strategy)

This strategy is nearly the same as strategy 5. Except for the first 2 steps.

The start square is at (2,3). I firstly set the first two moves as:

(2,3) -> (1,1) -> (3,2).

Then from (3,2), I used exactly the same approach of strategy 5. Moreover, each time when a solution found, I generated another solution which is the inverse path of this found solution. Since I have specified the direction of the first two steps, any generated solution will end with: (3,2) -> (1,1) -> (2,3). Thus any generated solution should be different from any of the program found ones. Moreover, since the program found ones are different with each other, the generated solutions should be different from each other as well. Therefore, all the solutions are distinct.

## Summary data

Chart 1: Number of moves per second for strategy 1-6. (S6 is my strategy)

Blue bars denote the mean values and red bars denote the median value. Error bars in black are the standard deviation.

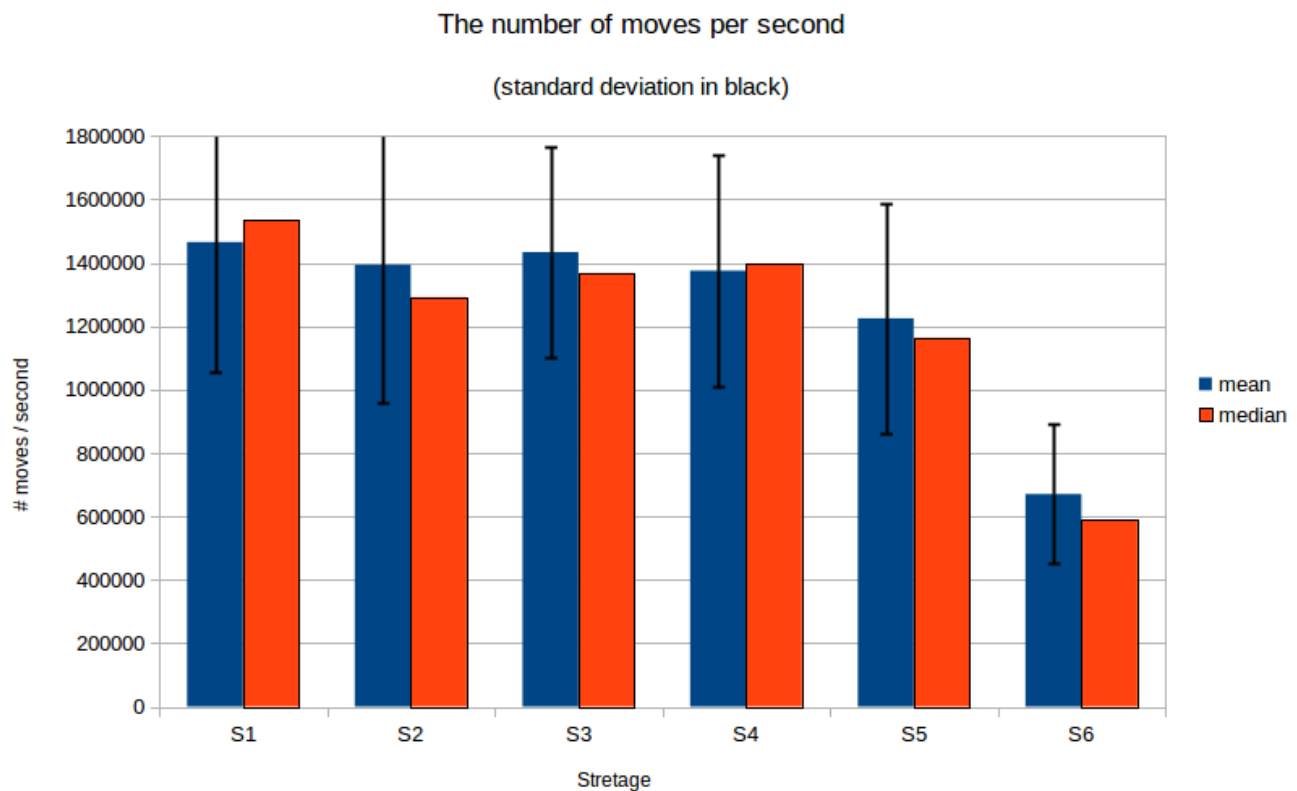


Chart 2: Number of sols per second for strategy 1-6. (S6 is my strategy)

Blue bars denote the mean values and red bars denote the median value. Error bars in black are the standard deviation.

## The number of solutions per second

(standard deviation in black)

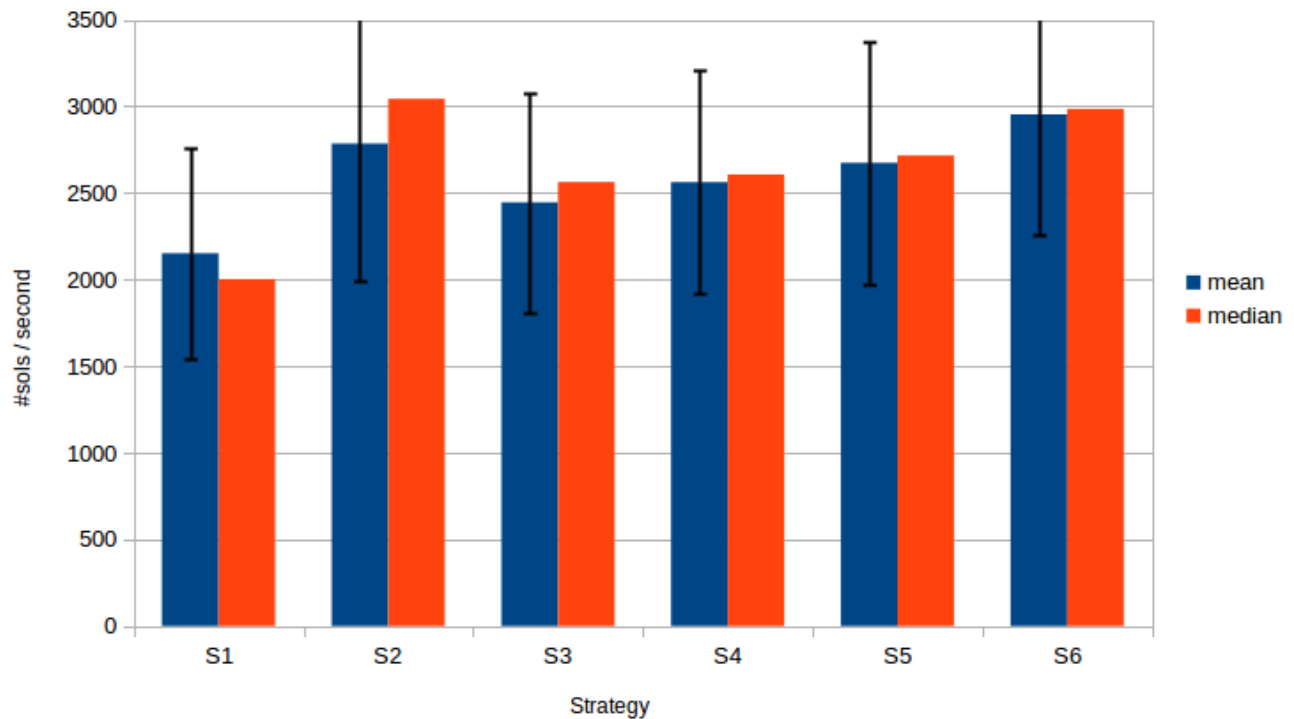


Table: Statistics result of 30 runs for each strategy:

| Num_moves/s    | Mean<br>Media<br>Std.Dev | S1           | S2           | S3           | S4           | S5           | S6           |
|----------------|--------------------------|--------------|--------------|--------------|--------------|--------------|--------------|
|                |                          | 1465194.2952 | 1393705.3408 | 1433393.8434 | 1375021.1119 | 1224758.9447 | 670362.93062 |
|                |                          | 1535404.1722 | 1291485.5688 | 1367658.154  | 1398318.1551 | 1163363.6761 | 589764.80673 |
|                |                          | 410953.34194 | 433789.61483 | 331869.08806 | 365990.6023  | 364133.53275 | 218775.97443 |
| Num_sols/sec   | Mean<br>Media<br>Std.Dev | S1           | S2           | S3           | S4           | S5           | S6           |
|                |                          | 2151.733744  | 2784.8697127 | 2445.8694049 | 2561.7512978 | 2674.3956398 | 2.95E+03     |
|                |                          | 2002.6493473 | 3043.7280799 | 2562.6527865 | 2606.7781029 | 2716.7047719 | 2984.6776151 |
|                |                          | 607.22760815 | 791.46085827 | 633.27801911 | 643.43442689 | 703.26388955 | 698.66752705 |
| Time elapse(s) | Mean<br>Media<br>Std.Dev | S1           | S2           | S3           | S4           | S5           | S6           |
|                |                          | 0.7470012904 | 0.7821249359 | 0.7332716015 | 0.7878138403 | 0.8843933027 | 1.6268337026 |
|                |                          | 0.6518226245 | 0.774568822  | 0.731181589  | 0.7151564085 | 0.859599609  | 1.696093633  |
|                |                          | 0.2450135103 | 0.2253356835 | 0.1650700015 | 0.2434199564 | 0.2504942087 | 0.447745429  |
| num of sols    | Mean<br>Media<br>Std.Dev | S1           | S2           | S3           | S4           | S5           | S6           |
|                |                          | 1651.4666667 | 2784.8697127 | 1886.3333333 | 2111.9333333 | 2493.5666667 | 5071.0666667 |
|                |                          | 1368.5       | 3043.7280799 | 1858.5       | 1904.5       | 2430.5       | 5269         |
|                |                          | 883.94783484 | 791.46085827 | 864.45184279 | 1060.6429388 | 1212.7893095 | 2245.5972853 |

### Analysis/takeaways from the graph:

Chart one shows that implementing more complicate heuristic will take more computation time for each move. However, by implementing heuristic to the searching strategy, it is more likely to get a

feasible solution and thus the time used to find each solution decreases. For example, strategy 1 uses the fixed degree as its heuristic and it doesn't need to update the degree for every move. Thus strategy 1 gets the most moves per second. However, this simple heuristic might be less likely than other strategies to find a feasible solution and thus it has the least sols per second.