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## INTRODUCTION

In this project, we are asked to develop an adversarial search agent to play the game “Isolation”. Isolation is a deterministic, 2-player game of perfect information in which the players alternate turns moving a single piece from one cell to another on a board. Rules are:

- Board is 7x7 grid
- Whenever either player occupies a cell, that cell becomes blocked for the remainder of the game
- This first player with no remaining legal moves loses, and the opponent is declared the winner.

Also, Udacity adds the following rules

- Agents will have a fixed time each turn to search for the best move and respond. If the time limits expire during a player's turn, that player forfeits the match, and the opponent wins.
- The agents can move to any open cell on the board that is 2-rows and 1-column or 2-columns and 1-row away from their current position on the board.

## HEURISTIC ANALYSIS

The goal is to develop a heuristic such that the coded agent (named Student) outperforms a base\_line agent called *ID improved*, that uses Iterative Deepening and the improved\_score heuristic defined in sample\_players.py provided.

The tournament.py script is used to evaluate the effectiveness of heuristic.

All the heuristics have been coded considering the relation between the possible legal moves for the two players (my player and the opponent players). Assuming that my player has to maximise the possible moves vs the opponents.

### HEURISTIC 1

The first heuristic is going to be squared difference between the number of the possible moves for each player.

$$\text{Heuristic 1} = \text{len}(\text{my possible moves})^2 - \text{len}(\text{opponent's possible moves})^2$$

### HEURISTIC 2

The second heuristic is going to be the difference between my movements and the opponents. However, I am going to weight the opponent's to 3 giving much more value to movements which are more distant. The weight has been set iterating some values.

$$\text{Heuristic 2} = \text{len}(\text{my moves}) - 3 * \text{len}(\text{opponent's possible moves})$$

### HEURISTIC 3

The third heuristic is going to be the Manhattan distance to the distance to the centre of the game.

$$\text{Heuristic 3} = \sum_{i=1}^n \text{Absolute}(\text{my distance to centre} - \text{opponent's distance to centre}_i)$$

## EVALUATION OF HEURISTICS

The script provided tournament.py is going to be used to measure the effectiveness of my custom heuristics. The script measures the relative performance of my *Student* agent a round-robin tournament against several other pre-defined agents. The *Student* agent uses time-limited Iterative Deepening along with my customs heuristics.

The tournament opponents are listed below

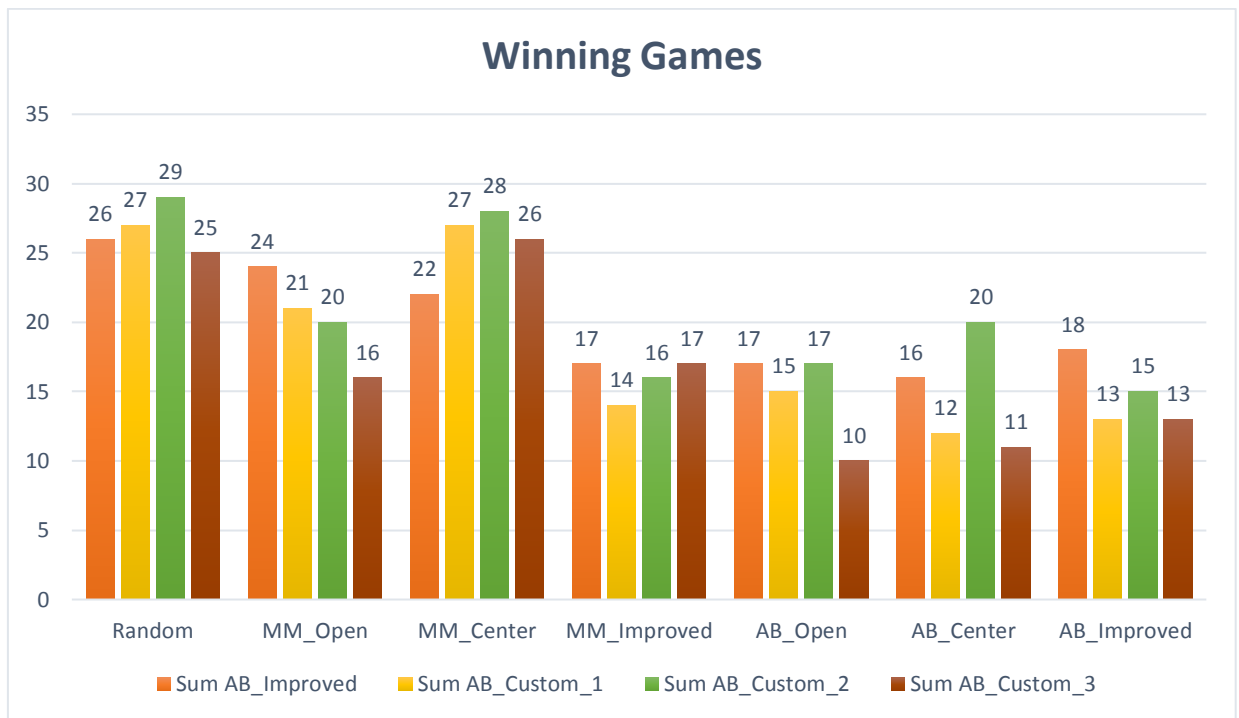
- **Random:** An agent that randomly chooses a move each turn.
- **MM\_Open:** MinimaxPlayer agent using the open\_move\_score heuristic with search depth 3
- **MM\_Center:** MinimaxPlayer agent using the center\_score heuristic with search depth 3
- **MM\_Improved:** MinimaxPlayer agent using the improved\_score heuristic with search depth 3
- **AB\_Open:** AlphaBetaPlayer using iterative deepening alpha-beta search and the open\_move\_score heuristic
- **AB\_Center:** AlphaBetaPlayer using iterative deepening alpha-beta search and the center\_score heuristic
- **AB\_Improved:** AlphaBetaPlayer using iterative deepening alpha-beta search and the improved\_score heuristic

## RESULTS

I did run the heuristics in 3 tournaments and summarised the game results in the following table:

Opponent	Sum AB_Improved		Sum AB_Custom_1		Sum AB_Custom_2		Sum AB_Custom_3	
	Won	Lost	Won	Lost	Won	Lost	Won	Lost
Random	26	4	27	3	29	1	25	5
MM_Open	24	6	21	9	20	10	16	14
MM_Center	22	8	27	3	28	2	26	4
MM_Improved	17	13	14	16	16	14	17	13
AB_Open	17	13	15	15	17	13	10	20
AB_Center	16	14	12	18	20	10	11	19
AB_Improved	18	12	13	17	15	15	13	17
Total general	140	70	129	81	145	65	118	92
%(win rate)	66.67%		61.43%		69.05%		56.19%	

And with plot:



My second heuristic (weighted difference on the number of movements) got better results on winning rate. Winning rate is 69% beating the AB\_improved (66,7%) in a bit more than two percentual points. Moreover, it hits the other heuristics in nearly all kind of heuristic opponent (4 out of 7).

The second best performance is for heuristic AB\_Improved which ranks better also in 4 out of 7 opponents but with less winning games.

## CONCLUSION

The heuristic chosen will be heuristic 2:

- It is a simple calculation and its definition it is plain and easy to interpret.
- It beats the majority of opponents (4 out of 7).
- It does not need to use another in-memory object. Therefore, the computation time is small.

## APPENDIX

***** Playing Matches *****									
Match #	Opponent	AB_Improved		AB_Custom		AB_Custom_2		AB_Custom_3	
		Won	Lost	Won	Lost	Won	Lost	Won	Lost
1	Random	7	3	9	1	9	1	10	0
2	MM_Open	6	4	7	3	6	4	6	4
3	MM_Center	7	3	10	0	9	1	8	2
4	MM_Improved	6	4	6	4	6	4	7	3
5	AB_Open	7	3	5	5	5	5	4	6
6	AB_Center	5	5	5	5	6	4	4	6
7	AB_Improved	7	3	6	4	4	6	6	4
Win Rate:		64.3%		68.6%		64.3%		64.3%	

***** Playing Matches *****									
Match #	Opponent	AB_Improved		AB_Custom		AB_Custom_2		AB_Custom_3	
		Won	Lost	Won	Lost	Won	Lost	Won	Lost
1	Random	10	0	8	2	10	0	8	2
2	MM_Open	9	1	8	2	6	4	5	5
3	MM_Center	9	1	10	0	10	0	9	1
4	MM_Improved	6	4	7	3	6	4	6	4
5	AB_Open	5	5	5	5	6	4	2	8
6	AB_Center	5	5	2	8	8	2	3	7
7	AB_Improved	5	5	3	7	6	4	4	6
Win Rate:		70.0%		61.4%		74.3%		52.9%	

***** Playing Matches *****									
Match #	Opponent	AB_Improved		AB_Custom		AB_Custom_2		AB_Custom_3	
		Won	Lost	Won	Lost	Won	Lost	Won	Lost
1	Random	9	1	10	0	10	0	7	3
2	MM_Open	9	1	6	4	8	2	5	5
3	MM_Center	6	4	7	3	9	1	9	1
4	MM_Improved	5	5	1	9	4	6	4	6
5	AB_Open	5	5	5	5	6	4	4	6
6	AB_Center	6	4	5	5	6	4	4	6
7	AB_Improved	6	4	3	7	5	5	3	7
Win Rate:		65.7%		52.9%		68.6%		51.4%	