Isolation

AIND-Project 2 Heuristic Evaluation

10/03/2017

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#analysis

Introduction:

This project is an introduction to several important concepts in artificial intelligence and machine learning. Several well known functions and models (alpha-beta, minimax, etc.) are implemented to show the results of decision tree searching evaluated with different criteria and parameters.

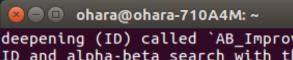
For this project, three different custom heuristics were implemented:

- 1) Movements based on center position [center]
- 2) Movements based on empty squares [empty]
- 3) Movements based on common sets [common]

Please note that more than three iterations were taken for testing, but these images best represent the data. Occasionally, heuristics would win all of the rounds in a testing phase and then lose all of the rounds in a subsequent test without altering any code. The averages in the tables represent average values when the sum of the results is normalized to the number of test completed.

Visual Representation:

Below are three different tables that were acquired by the terminal text-GUI. These images show three Tournament runs comparing the three custom heuristics to the improved AB function.



deepening (ID) called `AB_Improved`. The three `AB_Custom` agents use ID and alpha-beta search with the custom score functions defined in game_agent.py.

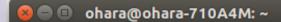
*****	*****
Playing	Matches
*******	******

Match #	Opponent	AB_Improved Won Lost		AB_Custom Won Lost		AB_Cus Won	tom_2 Lost	AB_Custom_3 Won Lost		
1	Random	9	1	9	1	7	3	9	1	
2	MM_Open	5	5	6	4	6	4	8	2	
3	MM_Center	6	4	6	4	8	2	10	0	
4	MM_Improved	5	5	7	3	6	4	6	4	
5	AB_0pen	5	5	5	5	5	5	6	4	
6	AB_Center	7	3	10	0	5	5	7	3	
7	AB_Improved	5	5	6	4	5	5	4	6	
	Win Rate:	60	. 0%	70	. ი%	60	0%	71	4%	

Your ID search forfeited 167.0 games while there were still legal moves availabl e to play.

(aind) ohara@ohara-710A4M:~\$

Test 2:



Win Rate:

This script evaluates the performance of the custom_score evaluation function against a baseline agent using alpha-beta search and iterative deepening (ID) called `AB_Improved`. The three `AB_Custom` agents use ID and alpha-beta search with the custom_score functions defined in game_agent.py.

Match #	Opponent	AB_Improved		AB_Custom		AB_Cu	stom_2	AB_Custom_3	
		Won	Lost	Won	Lost	Won	Lost	Won	Lost
1	Random	7	3	7	3	10	0	8	2
2	MM_Open	7	3	8	2	6	4	6	4
3	MM_Center	7	3	9	1	7	3	9	1
4	MM_Improved	6	4	7	3	3	7	5	5
5	AB_Open	5	5	8	2	4	6	5	5
6	AB_Center	4	6	7	3	5	5	7	3
7	AB_Improved	6	4	3	7	5	5	3	7

60.0%

Your ID search forfeited 168.0 games while there were still legal moves availabl

70.0%

57.1%

61.4%

Test 3:

Match #	Opponent	AB_Ir Won	nproved Lost	AB_C Won	ustom Lost	AB_Cus Won	tom_2 Lost	AB_Cus Won	tom_3
1	Random	7	3	9	1	8	2	10	0
2	MM_Open	6	4	7	3	7	3	8	2
3	MM_Center	7	3	10	0	6	4	8	2
4	MM_Improved	5	5	8	2	5	5	7	3
5	AB_Open	6	4	6	4	4	6	6	4
6	AB_Center	4	6	5	5	7	3	6	4
7	AB_Improved	4	6	7	3	6	4	5	5

Win Rate: 55.7% 74.3% 61.4% 71.4%

Your ID search forfeited 162.0 games while there were still legal moves available to play.

(aind) ohara@ohara-710A4M:~/Documents/Nanodegree/AIND/AIND-Isolation-master\$

#results

The results are that the [center] heuristic regularly has a higher performance than the other options, while the [common] heuristic is able to achieve a higher maximum W/L ratio. The [center] choice can usually dominate the Random, MM_improved and AB_Center tests, though it occasionally struggles against the AB_improved. The [common] heuristic can perform well against the MM_center test and has a robust performance elsewhere. The [empty] heuristic is usually on par with the AB_improved, but only performs slightly better. The results of the [empty] heuristic were unexpected, but it likely could be improved with some modifications.

Based on these results, the recommendation is to use the [center] heuristic, since it has a higher precision of results over multiple testing evaluations. While the [common] heuristic has a higher maximum W/L ratio, the difference between the resulting high value does not justify its usage.

- 1) Not all positions are equal. The center move provides optimal placement with the most freedom of choice for remaining moves.
- 2) Blocking can be achieved, since the board was partitioned and the next move can be the same as the opponent ("reflecting" in the lecture video).
- 3) If the move is the same as the opponent, an optimal move can be selected for the next move.

Final note: There were many forfeited moves based the code not being optimized. The heuristics would have a higher performance rating (W/L) if the search algorithms were optimized.