Artificial Intelligence Nanodegree - Project 2 - Build a Game-Playing Agent

Heuristic Analysis

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I experimented with quite a few heuristics. I found none that performed significantly better than the simple "own moves minus opponent moves" formula used by ID Improved. Though I never found a good combination that worked, the best solution probably would a combination of heuristics that uses one strategy early in the game and a different strategy late in the game. Also, to the extent possible, end-game specific logic could have a positive win rate boost.

Summary

Heuristic	Description	Win rate
Baseline	Own moves minus opponent moves (one level	67%
	deep)	
custom_score_A()	Standard own moves minus opponent moves for	70%
	most of game. End game uses second level moves	
Submitted code uses this	(own minus opponent) with a higher score for cases	
method.	where we can block one of the opponent's next	
	moves.	
custom_score_B()	Board position (inner, outer squares)	60%
custom_score_C()	Own moves minus opponent moves (second level	65%
	deep)	

Recommendation

Use custom_score_A() because:

- It uses different strategies for early game and end-game.
- It uses the simplistic but proven "own moves minus opponent moves" strategy for much of the game. This works when there are plenty of spaces.
- At end game it looks at econd-level moves and also assigns a higher score during the end-game when we have the opportunity to block the opponent's next move. This should give us an advantage.

Baseline – own moves minus opponent moves

Compare counts of next moves – one level deep.

Performance

Simple but tough to beat.

Playing Matches:

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```

```
Match 1: ID_Improved vs Random Result: 35 to 5
Match 2: ID_Improved vs MM_Null Result: 38 to 2
Match 3: ID_Improved vs MM_Open Result: 24 to 16
Match 4: ID_Improved vs MM_Improved Result: 23 to 17
Match 5: ID_Improved vs AB_Null Result: 32 to 8
Match 6: ID_Improved vs AB_Open Result: 21 to 19
Match 7: ID_Improved vs AB_Improved Result: 17 to 23
```

Results:

ID Improved 67.86%

custom_score_A() - multiple strategies

Standard "own moves minus opponent moves" for most of game. End game uses second level moves (own minus opponent). In the end game, we increase the score for cases where we can block one of the opponent's next moves.

Performance

In practice, this heuristic gave marginally better results than the baseline.

Playing Matches:

```
Match 1: Student vs Random Result: 38 to 2
Match 2: Student vs MM_Null Result: 36 to 4
Match 3: Student vs MM_Open Result: 23 to 17
Match 4: Student vs MM_Improved Result: 26 to 14
Match 5: Student vs AB_Null Result: 34 to 6
Match 6: Student vs AB_Open Result: 21 to 19
Match 7: Student vs AB_Improved Result: 18 to 22
```

Results:

Student 70.00%

custom_score_B() - board position

This heuristic assumes that inner board positions are advantageous over outer positions. In practice, there is probably much overlap with the standard "own moves minus opponent moves" in that inner positions also have more own moves.

We assume a 7x7 board and assign tiers, then assign a score to the current position (own and opponent):

- tier 0 = the ring of 24 outer squares
- tier 1 = the ring of 16 squares one space in from the outer ring
- tier 2 = the center 9 squares

Performance

The results were worse than the baseline.

```
Playing Matches:
```

```
Match 1: Student vs Random Result: 36 to 4
Match 2: Student vs MM_Null Result: 34 to 6
Match 3: Student vs MM_Open Result: 19 to 21
Match 4: Student vs MM_Improved Result: 16 to 24
Match 5: Student vs AB_Null Result: 29 to 11
Match 6: Student vs AB_Open Result: 20 to 20
Match 7: Student vs AB Improved Result: 14 to 26
```

Results:

Student 60.00%

custom_score_C() - own minus opponent - second level moves

Count the number of second moves (next move plus one) for each player. Use heuristic formula "own second moves minus opponent second moves." This should give a better sense of the future board score than the baseline.

Performance

In practice, this heuristic produced slightly worse results than the baseline.

Playing Matches:

```
Match 1: Student vs Random Result: 37 to 3
Match 2: Student vs MM_Null Result: 29 to 11
Match 3: Student vs MM_Open Result: 26 to 14
Match 4: Student vs MM_Improved Result: 22 to 18
Match 5: Student vs AB_Null Result: 33 to 7
Match 6: Student vs AB_Open Result: 19 to 21
Match 7: Student vs AB Improved Result: 18 to 22
```

Results:

Student 65.71%