

Heuristic Analysis

Rssult

```
Project2 Isolation — -bash — 80x24
~/Downloads/Artificial Intelligence/7P Advanced Game Playing/Project2 Isolation — -bash
dujinhong Project2 Isolation $ python tournament.py

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.

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

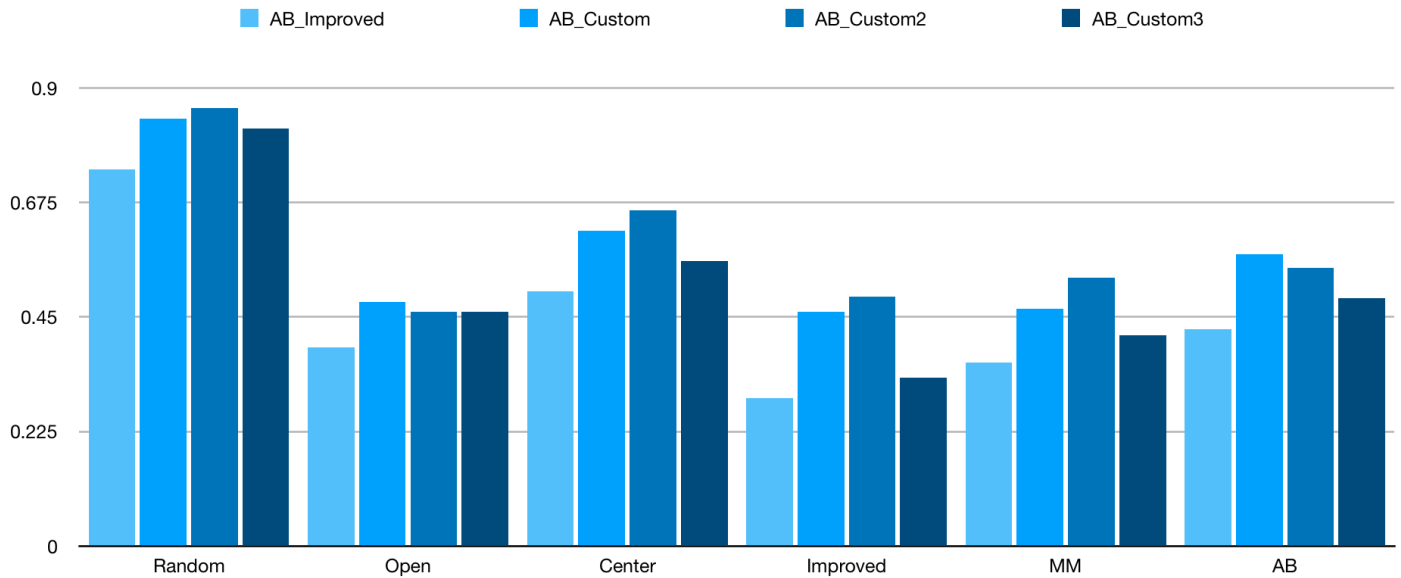
Match #   Opponent   AB_Improved   AB_Custom   AB_Custom_2   AB_Custom_3
              Won | Lost   Won | Lost   Won | Lost   Won | Lost
1         Random     37 | 13     42 |  8     43 |  7     41 |  9
2         MM_Open    16 | 34     21 | 29     25 | 25     20 | 30
3         MM_Center  28 | 22     34 | 16     33 | 17     31 | 19
4         MM_Improved 10 | 40     15 | 35     21 | 29     11 | 39
5         AB_Open    23 | 27     27 | 23     21 | 29     26 | 24
6         AB_Center  22 | 28     28 | 22     33 | 17     25 | 25
7         AB_Improved 19 | 31     31 | 19     28 | 22     22 | 28

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Win Rate:      44.3%      56.6%      58.3%      50.3%
dujinhong Project2 Isolation $
```

We can see the three heuristic function works better than the `AB_Improved` on almost time.

Result

	Random	MM_Open & AB_Open	MM_Center & AB_Center	MM_Improved & AB_Improved	MM	AB
AB_Improved	0.74	0.39	0.5	0.29	0.36	0.426666666666667
AB_Custom	0.84	0.48	0.62	0.46	0.466666666666667	0.573333333333333
AB_Custom2	0.86	0.46	0.66	0.49	0.526666666666667	0.546666666666667
AB_Custom3	0.82	0.46	0.56	0.33	0.413333333333333	0.486666666666667



Analysis

- Custom 1

Here I use the piecewise function, using different stages in different situations.

1. When one of the player has won, just return as follows.

```
if game.is_loser(player):
    return float("-inf")

if game.is_winner(player):
    return float("inf")
```

2. When the game just has begun for a short time,

$$\frac{2}{3} \leq \frac{\text{blank_spaces}}{\text{total_spaces}} \leq 1$$

I would like to force my player to conquer the center of the blank areas. It is because when there are many blank positions on the board, it will be more choices for the players on such areas instead of the edges of

the board. I use the following fomula to calculate the value of the heuristic function at this stage.

$$d = \|x - C\|_2^2$$

We can call it `improve_center_score`.

3. When the game proceeds for a while,

$$\frac{1}{3} < \frac{blank_spaces}{total_spaces} \leq \frac{2}{3}$$

I would like to combine the `improved_score` and the `improve_center_score`. Using

$$own_moves - 2 \times opp_moves + \frac{d}{total_spaces}$$

to be the value of the heuristic function at this stage. Here the second term plus by 2 because I want it to be more offensive, the third term divides by `total_spaces` because I want to reduce the influence of the `improve_center_score`, and I still keep it because when the former score give the same value on 2 different positions, the player can choose the better position.

We can call it `combined_improved_center_score`.

4. When the space of the board is almost run out,

$$0 < \frac{blank_spaces}{total_spaces} \leq \frac{1}{3}$$

the `improved_center_score` is no more work. Now I only want my players to survive and use the strategies of defend, so we only focus on the `open_move_score`

$$own_moves$$

```
center = np.mean(game.get_blank_spaces())
own_moves = len(game.get_legal_moves(player))
opp_moves = len(game.get_legal_moves(game.get_opponent(player)))
blank = len(game.get_blank_spaces())
if blank >= game.width*game.height/3*2:
    return np.sum((game.get_player_location(player) - center)**2)
elif blank >= game.width*game.height/3:
    return float(own_moves) - 1.5 * float(opp_moves) + np.sum((game.get_player_location(player) - center)**2)/game.width/game.height
else:
    return float(own_moves)
```

1. When one of the player has won, just return.

2. When

$$\frac{1}{2} \leq \frac{blank_spaces}{total_spaces} \leq 1$$

use the `combined_improved_center_score` to conquer best positions.

3. When

$$0 < \frac{blank_spaces}{total_spaces} < \frac{1}{2}$$

use

$$own_moves + \frac{d}{total_space}$$

as the value of the heuristic function at this stage.

* Custom 3

1. When one of the player has won, just return.

2. When

$$\frac{1}{2} \leq \frac{blank_spaces}{total_spaces} \leq 1$$

use the `combined_improved_center_score` to conquer best positions.

3. When

$$0 < \frac{blank_spaces}{total_spaces} < \frac{1}{2}$$

take a more offensive strategies and use

$$own_moves - 1.5 \times opp_moves$$

as the value of the heuristic function at this stage.

Recommendation

I will recommend to use the Custom 2 heuristic function for the following reasons.

- Win Rate

We can see Custom 1 and Custom 2 both work better than other functions. However, they are equally matched. But Custom 2 works better than Custom 1 in almost time.

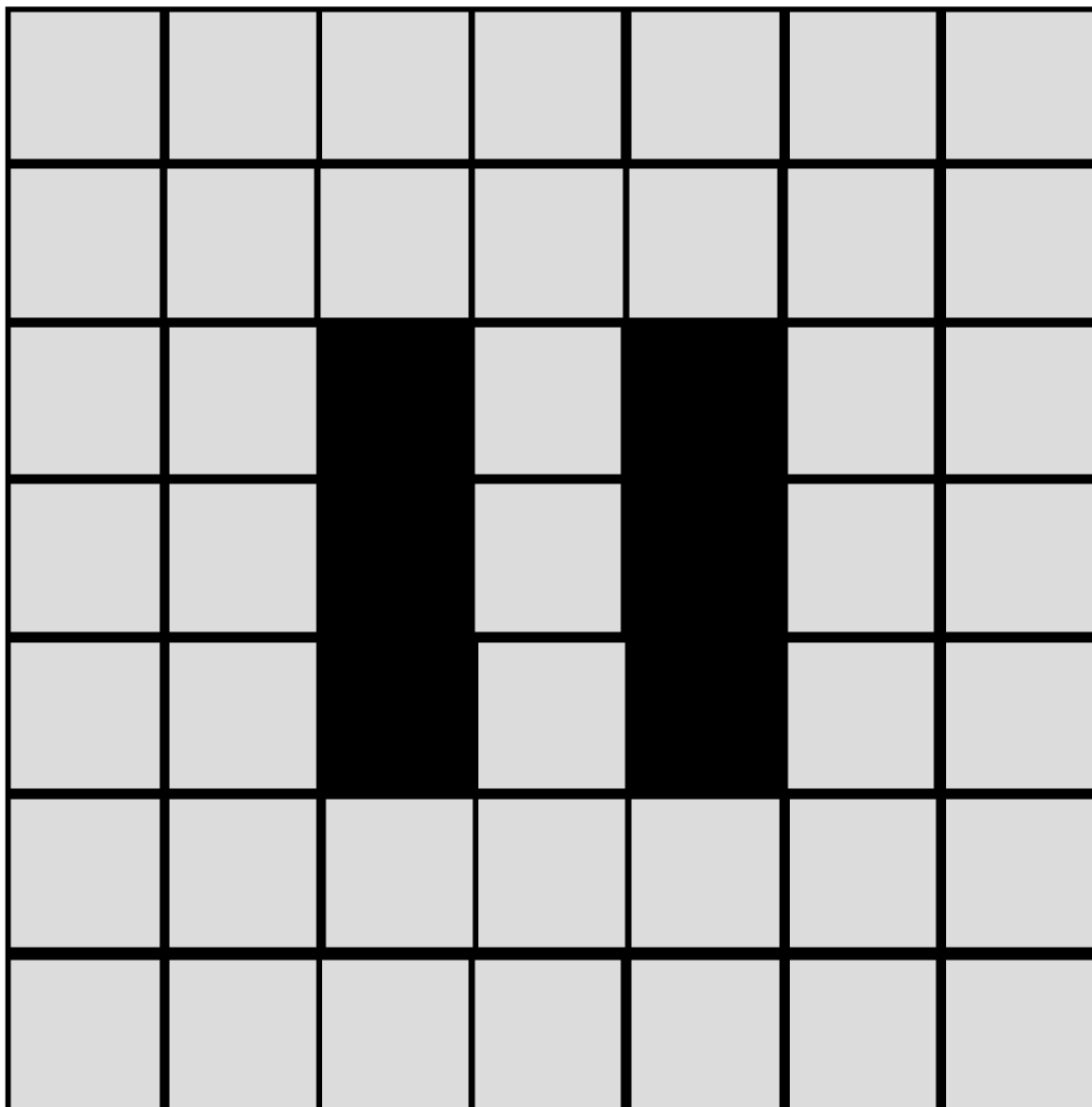
- Complexity

The Customs both have complexity $O(n)$ and Custom 2 has less calculation than Custom 1.

- How the Heuristic Predicts the Final Outcome of the Game

First, conquer the center of the blank areas.

Second, when the some blank areas have been blocked, the center-evaluation function help less. For example, the following situation will lead the center-evaluation function to value the positions near the center position to be better. However, it is better to stay outside.



Therefore, I will recommend to use Custom 2.