Artificial Intelligence Nanodegree Program

Project#3: Build an Adversarial Game Playing Agent

March 21st, 2022

(1) Game Agent Implementation

→ my_custom_player.py (minimax and alpha-beta search algorithm)

(2) Experimental Results & Report

(a)CustomAgent search function uses an advanced search technique

I chose Option 1: Develop a custom heuristic.

I implemented the following 3 heuristic functions.

(i) Difference between my moves and opponent's moves (default heuristic)

I implemented the default heuristic based on the lecture.

I also tuned the balance between my moves and opponent's moves.

```
# add score function

def score(self, state):

own_loc = state.locs[self.player_id]

opp_loc = state.locs[1 - self.player_id]

own_liberties = state.liberties(own_loc)

opp_liberties = state.liberties(opp_loc)

#return len(own_liberties) - len(opp_liberties) # heuristic 1 (original)

#return len(own_liberties) - 2 * len(opp_liberties) # heuristic 2

#return 2 * len(own_liberties) - len(opp_liberties) # heuristic 3

#return len(own_liberties) - 1.5 * len(opp_liberties) # heuristic 4

#return 1.5 * len(own_liberties) - len(opp_liberties) # heuristic 5

#return len(own_liberties) - 1.2 * len(opp_liberties) # heuristic 6

#return 1.1 * len(own_liberties) - len(opp_liberties) # heuristic 7

return 1.1 * len(own_liberties) - len(opp_liberties) # heuristic 8
```

(ii) How close from the center (original#1 heuristic)

Knight cannot move well near the corner, so I created an original heuristic that calculate how close from the center. I subtract it from the max distance: sqrt(4*4 + 5*5), so the bigger score the better result.

```
def score2(self, state):
    __WIDTH = 11
    owm_loc = state.locs[self.player_id]
    center_loc = 57
    x, y = own_loc % (_WIDTH + 2), own_loc // (_WIDTH + 2)
    cx, cy = center_loc % (_WIDTH + 2), center_loc // (_WIDTH + 2)
    center_distance = math.sqrt((cx-x)*(cx-x) + (cy-y)*(cy-y))
    return math.sqrt(41) - center_distance # how close from center point (farest point is 0) # heuristic 9
```

(iii) Combination of default heuristic and original heuristic (original#2 heuristic)

I combine the 2 heuristics by multiplying weights.

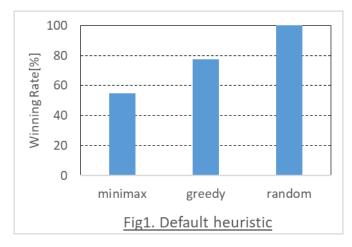
```
def score3(self, state):
    | _WIDTH = 11
    | own_loc = state.locs[self.player_id]
    | own_loc = state.locs[1 - self.player_id]
    | own_liberties = state.liberties(own_loc)
    | opp_liberties = state.liberties(opp_loc)
    | center_loc = 57
    | x, y = own_loc % (_WIDTH + 2), own_loc // (_WIDTH + 2)
    | center_distance = math.sqrt((cx-x)*(cx-x) + (cy-y)*(cy-y))
    | #return 1*(len(own_liberties) - len(opp_liberties)) + 1*(math.sqrt(41) - center_distance) # combination # heuristic 10
    | #return 2*(len(own_liberties) - len(opp_liberties)) + 2*(math.sqrt(41) - center_distance) # combination # heuristic 11
    | #return 1*(len(own_liberties) - len(opp_liberties)) + 2*(math.sqrt(41) - center_distance) # combination # heuristic 12
    | #return 1*(len(own_liberties) - len(opp_liberties)) + 3*(math.sqrt(41) - center_distance) # combination # heuristic 13
    | #return 1*(len(own_liberties) - len(opp_liberties)) + 3*(math.sqrt(41) - center_distance) # combination # heuristic 13
    | #return 1*(len(own_liberties) - len(opp_liberties)) + 3*(math.sqrt(41) - center_distance) # combination # heuristic 13
    | #return 1*(len(own_liberties) - len(opp_liberties)) + 3*(math.sqrt(41) - center_distance) # combination # heuristic 13
    | #return 1*(len(own_liberties) - len(opp_liberties)) + 3*(math.sqrt(41) - center_distance) # combination # heuristic 13
```

(b)Result

(i) Difference between my moves and opponent's moves (default heuristic)

I set the round = 10. At first, I checked the result of the default heuristic (minimax and alphabeta pruning) with <u>depth=3</u>, which is the same depth as the opponent agent.

My custom agent can win against random agent 100%, and also can win against greedy agent about 80%. However, It can win only 55% against minimax agent. Alpha-beta pruning can only improve the calculation efficiency, so it's understandable that the result is almost even against the minimax agent.

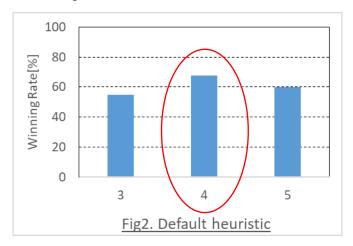


With the default heuristic, I also checked how depth change works.

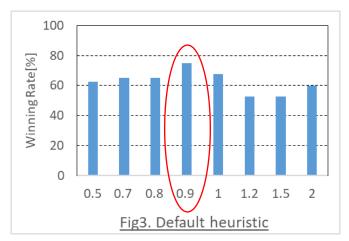
I fix the opponent as minimax agent here.

I can win more with the depth=4, but the result will saturate after depth=5.

I will take more calculation time with deeper depth, so <u>I choose depth=4</u> as the best depth in this project for the following calculations.



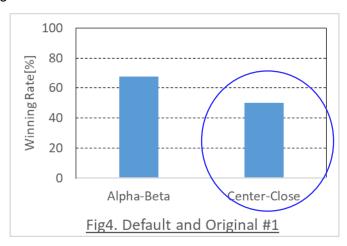
I changed the weight balance between my moves and opponent's moves and the result is shown below. X axis means the weight balance ratio. For example, the ratio of the heuristic "len(own_liberties) - 2 * len(opp_liberties)" is 2. It means that the bigger ratio, the more aggressive agent.



As the result, the ratio nearly equal 0.9: "1.1 * len(own_liberties) - len(opp_liberties)", a little less aggressive agent, has the <u>best result: 75%.</u>

(ii) How close from the center (original#1 heuristic)

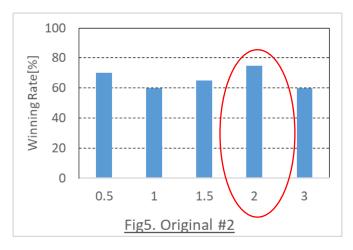
Only using my original#1 heuristic: center-close doesn't work so well as below.



(iii) Combination of default heuristic and original heuristic (original#2 heuristic)

I changed the weight balance between default heuristic (my moves and opponent's moves) and original#1 heuristic.

The result is shown below. X axis means the weight balance ratio. For example, the ratio of the heuristic "1*(len(own_liberties) - len(opp_liberties)) + 2*(math.sqrt(41) - center_distance)" is 2. It means that the bigger ratio, the more center-close agent.



As the result, the ratio equal 2 : "1*(len(own_liberties) - len(opp_liberties)) + 2*(math.sqrt(41) - center_distance)" has the <u>best result: 75%.</u>

(c)Answer to the rubric questions

The answers of the Option1: Advanced Heuristic are show below.

- What features of the game does your heuristic incorporate, and why do you think those features matter in evaluating states during search?

I incorporated the combination of the following 2 features in my heuristic.

- (i)Subtraction of my moves and opponent's moves
- (ii)How close to the center
- (i) is the default heuristic, and it's intuitively understandable to maximize the winning probability.
- (ii) is my original heuristic, and it's intuitively better not to go near to the board edges because the movement of Knight would be restricted.

I thought it's better to combine these 2 heuristics and give a good balance weight by trying some couples. As a result, I got a good answer.

- Analyze the search depth your agent achieves using your custom heuristic. Does search speed matter more or less than accuracy to the performance of your heuristic?

I changed the depth from 3 to 5 with the final setting.

Same as Fig.2 above, the winning rate will be saturated after the depth=5.

It's much slower than depth=4, so I think <u>depth=4 is the best</u> in this search problem.

