

Heuristic analysis

Udacity - Artificial Intelligence Nanodegree

- custom_score1

In this function, I calculated the distance from center. Because I thought there are a lot of legal move cases in the center. I thought it could make more efficient move. The closer to the center, The player will get high score. So contrary to other scores, I set minus to `own_distance_from_center`. And I also set plus to `opp_distance_from_center`. Because those values mean distance from center.

$$\text{-own_distance_from_center + opp_distance_from_center + own_moves - opp_moves}$$

In this case, the agent will get reward And then, added basic score. And, when a player lost the game, the agent will get negative reward ($-\infty$). Contrary, when a player win the game, the agent will get positive reward(∞).

But, I wasn't as efficient as I thought. It was very similar to `AB_improved`. Because opponent's algorithm also find center position. So, each distance from center wasn't significant difference. So it was almost same as `AB_improved`.

- custom_score2

This function is very similar with '`AB_improved`'. I added *blank_spaces*. The Blank spaces means the locations that are still available on the board. So I thought it can helpful for our score. Except this, all of things are same as `AB_improved`.

$$\text{blank_spaces + own_moves - opp_moves}$$

In my case, it was efficient. In previous score, I expected more space make more chance to win. It's correct. But, It doesn't means the player always locate in center position. As the game progresses, center position couldn't guarantee more legal movements. It depend on game situation. So, I think calculating blank spaces is more efficient than calculating distance from center.

- custom_score 3

In this case, I calculated same spaces in legal moves between own player and opponent player. I thought occupying same legal space first can make reduced opponent's legal moves number. It can remove opponent player's movements. So I thought it can be efficient measure.

own_moves - opp_moves + similar_count

Actually I thought score 3 is better than others. But, score 2 was best. So, I understood there is not perfect solution. It depended on situations. And, I think it isn't efficient for big isolation game. Because it has loop part.

- Conclusion

***** Playing Matches *****									
Match #	Opponent	AB_Improved		AB_Custom		AB_Custom_2		AB_Custom_3	
		Won	Lost	Won	Lost	Won	Lost	Won	Lost
1	Random	15	5	16	4	18	2	16	4
2	MM_Open	12	8	11	9	14	6	10	10
3	MM_Center	16	4	13	7	13	7	16	4
4	MM_Improved	13	7	12	8	11	9	14	6
5	AB_Open	11	9	11	9	11	9	9	11
6	AB_Center	10	10	12	8	11	9	10	10
7	AB_Improved	10	10	12	8	10	10	12	8
Win Rate:		61.0%		62.1%		62.9%		61.4%	
***** Playing Matches *****									
Match #	Opponent	AB_Improved		AB_Custom		AB_Custom_2		AB_Custom_3	
		Won	Lost	Won	Lost	Won	Lost	Won	Lost
1	Random	14	6	10	2	10	2	16	4
2	MM_Open	11	9	13	7	16	4	13	7
3	MM_Center	16	4	11	9	17	3	14	6
4	MM_Improved	11	9	13	7	14	6	10	10
5	AB_Open	8	12	11	9	11	9	11	9
6	AB_Center	9	11	9	11	12	8	13	7
7	AB_Improved	10	10	10	10	0	12	10	10
Win Rate:		66.4%		60.7%		68.6%		62.1%	
***** Playing Matches *****									
Match #	Opponent	AB_Improved		AB_Custom		AB_Custom_2		AB_Custom_3	
		Won	Lost	Won	Lost	Won	Lost	Won	Lost
1	Random	19	1	17	3	17	3	17	3
2	MM_Open	14	6	14	6	15	5	13	7
3	MM_Center	14	6	13	7	13	7	16	4
4	MM_Improved	11	9	11	8	14	6	13	7
5	AB_Open	9	11	11	8	13	7	12	8
6	AB_Center	11	9	11	9	12	8	9	11
7	AB_Improved	13	7	9	11	9	11	9	11
Win Rate:		66.0%		62.9%		55.4%		63.6%	

I set number of matches in tournament.py as 10. And the following table is that summarized result.

	AB_Improved	distance from center	blank spaces	similar moves
Run 1	59.2%	59.2%	58.3%	58.3%
Run 2	54.1%	55.8%	65.0%	59.2%
Run 3	60.0%	59.2%	63.3%	60.0%
Average	57.8%	58.1%	62.2%	59.2%

In random algorithm, all of the scores are quite well. But in Minimax or Alpha-beta pruning, little different. Here is the result except random situation.

	AB_Improved	distance from center	blank spaces	similar moves
Run 1	61.4%	62.1%	62.9%	61.4%
Run 2	56.4%	60.7%	68.6%	62.1%
Run 3	65.0%	62.9%	66.4%	63.6%
Average	60.9%	61.9%	66.0%	62.4%

So, I choose custom_score 2: added black spaces. Because

1. Only custom_score 2 has over 60.0% average. Sometimes it had poor score than others. But, generally it was best.
2. This score is easy to understand and implement. Just
3. It just check current state. It is efficient because some algorithm should check previous state. Then it can cause lack of memory or overhead.