## **Heuristic Analysis**

## **Approach**

In order to devise the *custom\_score* functions logic, the following approach was used:

- 1. Using self-observation and intuition when playing the game of isolation, I wrote down a set of strategies I noticed myself applying during the game to evaluate the state of my player in the game at any point in time. The following, are those strategies:
  - Comparing the amount of my own moves vs the opponent's: the more moves I had over my opponent's, the better.
  - My ability to block the opponent on their next move: since the goal of the game is to isolate the opponent, blocking their future moves seemed like a natural strategy.
  - My ability to take over the center of the board on my next move: this was also pointed out during the class instructional videos.
  - A relative measure of how close my player was to the center as compared to the opponent: in general, if my player was closer to the center than the opponent, the better it seemed, as it creates the notion of "controling" the center.
  - A simple count of the number of possible moves for my player: this is also captured in the out-of-the-box open-move strategy.

There were other strategies I came up with during my brainstorming, but they were later discarded as irrelevant after further evaluation.

2. All of these strategies were later coded individually or in combination in the *custom\_score* functions available in <code>game\_agent.py</code>. When coded in combination, a similar approach to the one used when creating a Machine Learning model was utilized. I.e., a score function composed of the summation of the weighted features (strategies in this case) was applied. The general equation of the resulting function is:

$$score = \sum_{i=1}^{n} \omega_i * x_i$$
 , where  $n = number of strategies$ 

- 3. Afterwards, their performance were evaluated by running tournament.py over sevaral runs. The evaluation included observing the following characteristics in the results:
  - noting the total win rate for each test player.
  - comparing all AB\_Custom\* players to each opponent's result (paying attention to the opponent's strategy).
  - comparing all AB\_Custom\*\* players results to the AB\_Improved\* test player's for each opponent.
- 4. In an iterative fashion, each  $custom\_score$  function was updated, by changing their param weights  $(\omega_i)$  and adding or removing features to the equation (x) and further evaluated. This process allowed for the fine-tuning and optimization of the functions.

5. After several iterations, the final version of the *custom\_score* functions was produced. As the last verification step, a *tournament* consisting of "NUM\_MATCHES = 1000" was executed. The following screen shot summarizes the final results.

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Match #	Opponent	AB_Improved		AB_Custom		AB_Custom_2		AB_Custom_3	
		Won   L	.ost	Won	Lost	Won	Lost	Won	Lost
1	Random	1869	131	1862	138	1889	111	1862	138
2	MM_Open	1083	917	1117	883	1099	901	1022	978
3	MM_Center	1762	238	1798	202	1790	210	1733	267
4	MM_Improved	1027	973	1033	967	1035	965	961	1039
5	AB_Open	1044	956	1102	898	1060	940	1026	974
6	AB_Center	1707	293	1753	247	1747	253	1705	295
7	AB_Improved	1021	979	1078	922	1029	971	953	1047
	Win Rate:	68.0%		69.6%		68.9%		66.2%	

<sup>\*</sup>Results of tournament final version of the custom\_score functions over 1000 matches\*

The results show both  $AB\_Custom$  and  $AB\_Custom\_2$  perform slightly better than the  $AB\_Improved$  test agent. Also, it is interesting to note that the strategy implemented by  $AB\_Custom$  (custom\_score()) tends to perform really well against the out-of-the-box "center" strategy (as shown by the performance against  $MM\_Center$  and  $AB\_Center$ ); however, not as well, yet still better, when compared against the "open-move" and the "improved" strategies.

This suggests that certain strategies work better than others under certain situations, depending on what the opponent's strategy is. In other words, there may not be a silver-bullet strategy that works well on all situations. A better, and more intuitive approach is to use an adaptive stragegy - one that can dynamically adapt to the opponent's behavior, much like a smart player would act.

In conclusion, it is recommended that the AB\_Custom evaluation function be used, because:

- a) It performs better than the rest of the test agents under all provided scenarios.
- b) It is relatively simple, as it only combines 2 basic strategies (as described below).
- c) It can easily be fine-tuned, since it only has 2 features.

## Strategies utilized per custom\_score function

The following are the strategies utilized for each *custom\_score* function:

- custom score():
  - opponent\_block\_ability
  - relative mobility
- custom\_score\_2():
  - center\_ability
  - opponent\_block\_ability

- own\_mobility
- relative\_center\_domination
- relative\_mobility
- custom\_score\_3():
  - opponent\_block\_ability
  - own\_mobility

Name	Strategy				
center_ability	Ability to capture board center				
opponent_block_ability	Ability to block the opponent				
own_mobility	Player # of next moves				
relative_center_domination	Player's distance to center vs opponent's				
relative_mobility	Player's moves vs opponent's				

Strategy names as used in the \*custom\_score\* functions