

I tried 4 different heuristics –

- Custom Score #1: Return negative of number of opponent moves
  - Here, we are evaluating the best move by finding the move that leads to least number of opponent moves
- Custom Score #2: Return Number of Player moves – 2\*No of Opponent moves
  - Here, we are evaluating the best move by finding the move that leads to better moves for the current player than the opponent player
- Custom Score #3: Return the difference between current and opponent Player's distance from the board center
  - Here, we are evaluating the best move by finding the move that brings the current player closer to the center than the opponent.
- Custom Score #4: Combination of score logic from #2 & #3
  - Here, we are evaluating the best move by using the combined logic from points 2 & 3 above.

#### Comparison between Custom Score logic 1, 2 & 3:

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Playing Matches									
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Match #	Opponent	AB_Improved		AB_Custom		AB_Custom_2		AB_Custom_3	
		Won	Lost	Won	Lost	Won	Lost	Won	Lost
1	Random	10	0	7	3	8	2	9	1
2	MM_Open	7	3	7	3	8	2	5	5
3	MM_Center	8	2	9	1	8	2	6	4
4	MM_Improved	10	0	8	2	9	1	8	2
5	AB_Open	6	4	5	5	7	3	5	5
6	AB_Center	6	4	6	4	7	3	6	4
7	AB_Improved	4	6	4	6	5	5	4	6
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Win Rate:		72.9%		65.7%		74.3%		61.4%	

Your ID search forfeited 246.0 games while there were still legal moves available to play.

#### Comparison between Custom score logic 4, 2 & 3:

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 Playing Matches  
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Match #	Opponent	AB_Improved	AB_Custom	AB_Custom_2	AB_Custom_3
		Won   Lost	Won   Lost	Won   Lost	Won   Lost
1	Random	8   2	8   2	8   2	7   3
2	MM_Open	6   4	6   4	7   3	7   3
3	MM_Center	9   1	8   2	9   1	9   1
4	MM_Improved	9   1	7   3	8   2	8   2
5	AB_Open	7   3	4   6	6   4	2   8
6	AB_Center	10   0	7   3	7   3	7   3
7	AB_Improved	7   3	5   5	6   4	4   6
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Win Rate:		80.0%	64.3%	72.9%	62.9%

Your ID search forfeited 249.0 games while there were still legal moves available to play.

### Observations & Analysis:

From all the scoring functions that I tried, custom scoring function #2 (“Number of Player moves – 2\*No of Opponent moves”) gives the best performance since it compares the available options of both the current player and the opponent. Custom Score logic #4 combines the best of #2 & #3 but is less effective (than #2) because its computation logic takes much longer resulting in exploration of fewer levels (and nodes by iterative deepening).

### Recommendations:

Based on my analysis, I would like to recommend my evaluation function #2 which is “my\_moves – 2\*my\_opponent\_moves”. The main reasons are –

- Better performance (win rate) compared to the other evaluation functions that I tested
- This weighs the opponents\_moves higher than the current player’s moves for more aggressive game play and chases the opponent
- The timeouts/forfeits isn’t bad with this approach and it compares well with other approaches. I updated tournament.py to display the timeouts & forfeits for my evaluation functions 1, 2 & 3. Here are the observations (didn’t adjust headers & footer rows)

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 Playing Matches  
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Match #	Opponent	AB_Improved	AB_Custom	AB_Custom_2	AB_Custom_3
		Won   Lost   Tout   Ffeit	Won   Lost   Tout   Ffeit	Won   Lost   Tout   Ffeit	Won   Lost   Tout   Ffeit
1	Random	10   0    0   7	7   3    0   7	7   3    0   7	9   1    0   7
2	MM_Open	8   2    0   40	8   2    0   40	9   1    0   40	6   4    0   40
3	MM_Center	8   2    0   40	9   1    0   40	6   4    0   40	8   2    0   40
4	MM_Improved	10   0    0   40	8   2    0   40	9   1    0   40	9   1    0   40
5	AB_Open	5   5    0   40	2   8    0   40	5   5    0   40	6   4    0   40
6	AB_Center	5   5    0   40	8   2    0   40	8   2    0   40	5   5    0   40
7	AB_Improved	3   7    0   40	5   5    0   40	8   2    0   40	6   4    0   40
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Win Rate:		70.0%	67.1%	74.3%	70.0%