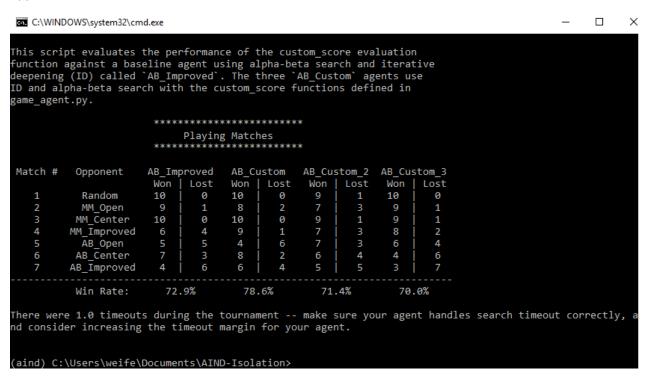
Analysis of Heuristic Evaluation Functions for Game-Playing Agent

Three heuristic evaluation functions were created to estimate the chance of winning when the agent has to cut off the search due to time limit.

- Function 1: # of open moves of the player 1.2 * # of open moves of the opponent + distance to center
- Function 2: # of unique open moves of the player 1.2 * # of unique open moves of the opponent
- Function 3: # of open moves of the player 1.2 * # of open moves of the opponent

Three agents using the above three heuristic evaluation functions, together with a baseline agent using the alpha-beta search and iterative deepening algorithm AB_Improved, were played against seven opponents. Tournament results are shown in the screen shot below:



Overall, Function 1 has the highest win rate: 78.6%, followed by the default agent AB_Improved (72.9%), Function 2 (71.40%), and Function 3 (70.0%). Therefore, I would choose Function 1. Below are three reasons I chose Function 1:

- 1. Function 1 has the highest winning rate in multiple tournaments
- 2. Although this function needs to calculate three terms, they are very easy to calculate. This means it should not affect search time at each level too much and therefore should not prevent it from going deeper levels compared to other algorithms

3.	I believe Function 1 is closer to the optimal evaluation function because it takes into account three factors which are essential to game winning. Therefore, this function is expected to perform better with a wider variety of opponent algorithms