Heuristics Analysis

Three heuristics are described in this summary. Following there is a description of each heuristic and the results obtained against the opponent agents (Random, MM_Null, MM_Open ,MM_Improved ,AB_Null ,AB_Open, AB_Improved) and the challenger agent (ID_Improved). The goal is to maximize the value returned by the heuristic for the maximizing player and minimize its value for the opponent.

1) Positions closer to the center are preferred since they give more freedom to the player

This heuristic is implemented using euclidean distance based on the center of the board and the position of the player for the current board state. The return value can be mathematically represented using the following expression

$$-\sqrt{(position(x) + 1 - \frac{board.height}{2})^2 + (position(y) + 1 - \frac{board.width}{2})^2)}$$

2) Positions closer to the borders should be avoided since they give less freedom to the player

This heuristic takes the minimum distance to any of the borders of the border and tries to maximize it to use the most distant position its 4 borders. The following expression illustrates the heuristic:

$$min(pos(x)+1, board.height-(pos(x)+1), pos(y)+1, board.width-(pos(y)+1))$$

3) Finally, a combination of these two plus number of available moves is used in a finally heuristic which is matched agains the challenger agent (ID_IMPROVED). Here is the expression:

$$-\sqrt{(position(x)+1-\frac{board.height}{2})^2+(position(y)+1-\frac{board.width}{2})^2)}+legalMoves()$$

Results

Setting: 100 games against each opponent. Results are the ones shown by the tournament.py script.

1) This heuristic gives poor results. The student agent is almost 10% below ID_Improved. All of the opponent players are using fixed depth search and the student is not able to win in the total against MM_Improved (50%) and AB_Open (48%). As expected for both test agents (ID_Improved and student) it is more difficult to win against an agent with a more elaborated heuristic (such as Improved) than versus one with a simpler expression (Null) for both search techniques (Minimax and alpha beta pruning).

Heuristic: Avoid moves closer to the edge			
Evaluating	ID_Improved	Student	
Opponent	Winning rate (%)	Winning rate (%)	
Random	83	77	
MM_Null	78	60	
MM_Open	65	64	
MM_Improved	61	50	
AB_Null	69	58	
AB_Open	64	48	
AB_Improved	59	56	
Results	68.43	59	

2) This heuristic offers a better match against the Improved one. However the student agent is behind by 3% approximately. Again the result versus all the opponents are as expected with the exception of ID_Improved winning with a better winning rate versus AB_OPEN (73%) than versus AB_NULL (68%). Probably a higher number of matches would allow us to see a clear difference in this regard.

Heuristic: moves closer to the center of the board are better			
Evaluating:	ID_Improved	Student	
Opponent	Winning rate (%)	Winning rate (%)	
Random	79	78	
MM_Null	77	70	
MM_Open	67	68	
MM_Improved	56	58	
AB_Null	68	68	
AB_Open	73	58	
AB_Improved	60	57	
Results	68.57	65.29	

3) This is the heuristic with the best results. A combination of favoring those moves closer to the center of the board and maximizing the number of allowed moves given the current state. One interesting observation is that the result of wins for ID_Improved was quite similar in all three tests, which indicates that the intention of having a hardware agnostic base reference of performance to compare with the student agent may be working. The result is clear, this heuristic beats the improved one. Random moves loses most of the matches against both test agents. And most of the times Null performs worse than Open, and Open performs worse than Improved. And since all of them are using fixed depth search, they perform worse against variants using iterative deepening.

Heuristic: moves closer to the center of the board are better and maximize number of legal moves			
Evaluating:	ID_Improved	Student	
Opponent	Winning rate (%)	Winning rate (%)	
Random	83	86	
MM_Null	73	78	
MM_Open	74	68	
MM_Improved	62	65	
AB_Null	69	78	
AB_Open	57	55	
AB_Improved	59	62	
Results	68.14	70.29	