

Heuristic Analysis

Introduction

The analysed position evaluation functions are all a variation of counting the number of legal moves minus the number of legal moves available to the opponent:

$$\text{Count}(\text{Moves}) - \text{Count}(\text{OpponentMoves})$$

Realising that some moves are more valuable than others one can sum the value of each move instead of just counting the moves. Subsequently the evaluation function was defined as:

$$\begin{aligned} & \text{Sum}(\text{Map}(n \rightarrow \text{Pow}(\text{Value}(n), N)), \text{Moves})) \\ & - \text{Weight} * \text{Sum}(\text{Map}(n \rightarrow \text{Pow}(\text{Value}(n), N)), \text{OpponentMoves}) \end{aligned}$$

whereas:

Value	Defines the value of single move
Pow	Raises the value to the power of N to give more weight to valuable moves
Weight	Is an additional factor that helps eliminating especially valueable opponent's moves

Value of Move

The value of a move is defined as the number of moves available on an empty board from a certain position. These numbers on a 7x7 board are:

Row/Column	0	1	2	3	4	5	6
0	2	3	4	4	4	3	2
1	3	4	6	6	4	4	3
2	4	6	8	8	8	6	4
3	4	6	8	8	8	6	4
4	4	6	8	8	8	6	4
5	3	4	6	6	4	4	3
6	2	3	4	4	4	3	2

These numbers were then normalized to values as shown below:

Row/Column	0	1	2	3	4	5	6
0	0.25	0.38	0.5	0.5	0.5	0.38	0.25
1	0.38	0.5	0.75	0.75	0.5	0.5	0.38
2	0.5	0.75	1	1	1	0.75	0.5
3	0.5	0.75	1	1	1	0.75	0.5
4	0.5	0.75	1	1	1	0.75	0.5
5	0.38	0.5	0.75	0.75	0.5	0.5	0.38
6	0.25	0.38	0.5	0.5	0.5	0.38	0.25

Variations

To test the idea following instances of the evaluation functions were defined and analysed:

Functionnames	N = 1	N = 2 (squared)	N = 3(cubed)
Weight = 1	ValueScore	Value2Score	Value3Score
Weight = 2	ValueWeightedScore	Value2WeightedScore	Value3WeightedScore

For example, function `Value2WeightedScore` squares all `Values` and applies a weight of 2.

Analysis

The above defined evaluation functions were tested using the Script *tournament.py*. The functions were then compared to the function AB_Improved which served as a benchmark. 10 rounds were played using the functions ValueScore, Value2Score and Value3Score. Another 10 rounds were played using the functions ValueWeightedScore, Value2WeightedScore and Value3WeightedScore. The “Win Rate” of a certain function was divided by the “Win Rate” of the benchmark function AB_Improved. Following table shows the results:

Algorithm / Tournament	Rnd 1	Rnd 2	Rnd 3	Rnd 4	Rnd 5	Rnd 6	Rnd 7	Rnd 8	Rnd 9	Rnd 10
AB_Improved	1.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00
ValueScore	1.13	1.02	0.94	0.91	1.22	0.94	0.96	1.02	0.96	0.98
Value2Score	1.07	1.12	1.02	0.96	1.11	1.00	1.09	1.00	1.12	0.92
Value3Score	1.09	0.96	1.06	0.92	1.16	1.00	1.06	1.02	0.96	0.87

Algorithm / Tournament	Rnd 1	Rnd 2	Rnd 3	Rnd 4	Rnd 5	Rnd 6	Rnd 7	Rnd 8	Rnd 9	Rnd 10
AB_Improved	1.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00
ValueWeightedScore	1.00	1.14	1.21	0.94	1.03	1.00	0.90	1.06	0.90	1.02
Value2WeightedScore	1.02	1.06	1.28	1.08	1.00	1.04	1.00	1.04	1.04	0.96
Value3WeightedScore	0.96	0.98	1.19	0.92	1.06	1.04	0.98	1.00	1.08	0.98
	Best in Round									
	Worst in Round									

Following table summarises the “Win Rates” and algorithms chosen:

Ranking	Algorithm / Tournament	Nr	Nr	Nr Best/Nr Worst	Nr > 1
	A	Best	Worst		
2.)	AB_Improved	2	2	1	0
	ValueScore	2	5	0.4	4
	Value2Score	3	0	-	6
	Value3Score	1	2	0.5	5
	Algorithm / Tournament				
	B				
	AB_Improved	0	2	0	0
3.)	ValueWeightedScore	3	3	1	5
1.)	<u>Value2WeightedScore</u>	<u>4</u>	<u>1</u>	<u>4</u>	<u>7</u>
	Value3WeightedScore	3	3	1	4

Best function was “Value2WeightedScore” which was 4 out of 10 rounds the best algorithm and only once the worst. It was 7 out of 8 times better than AB_Improved.