

Heuristic Analysis

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The tournament test consists of rounds of 5 matches each with a 150 milliseconds timeout. The results were compared to AB_Improved as baseline.

The x-axis correspond chosen custom heuristics:

1. AB_Improved
2. AB_Custom
3. AB_Custom_2
4. AB_Custom_3

Given opponent to the the number of possible moves to the adversary player, and player the number or moves of my agent, the custom heuristics score are given by:

1. $\text{player} - 1.618033988749895 * \text{opponent}$
2. $\text{player} - 2 * \text{opponent}$
3. $\text{player} / \text{opponent}$

Ten tournaments were executed, with 5 matches each, obtaining the following win rate percentual:

	win rate (% mean)	Win rate (% std)
AB_Improved	75.57	2.55
AB_Custom	77.43	3.68
AB_Custom_2	78.14	3.56
AB_Custom_3	78.14	3.86

Table 1. Average win rate

One might say that AB_Custom_2 and AB_Custom_3 are the winner heuristics, but in fact all algorithms are even.

As we can see in the candestick chart, all measurements overlap if we consider each standard deviation measurement, thus due the current sampling number if is not possible to define a winner strategy.

Future improvements to all strategies we should consider to enlarge the gap between them to AB_Improved are:

- Use multiprocessing or multithreading to calculate each branch of the moves possibility tree in parallel.
- Create opening moves, such as we previously did on AIND-Sudoku

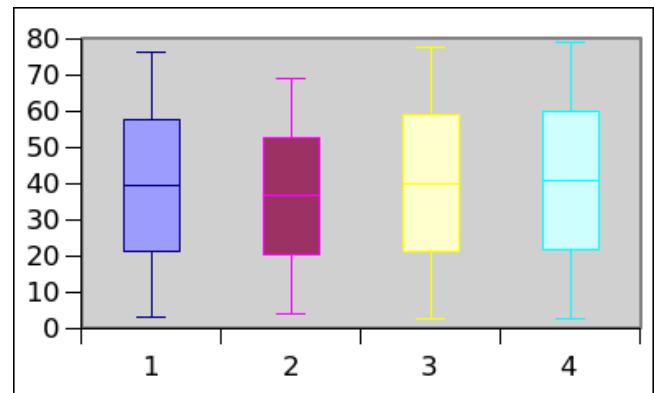


Figure 1. Average win rate with standard deviation

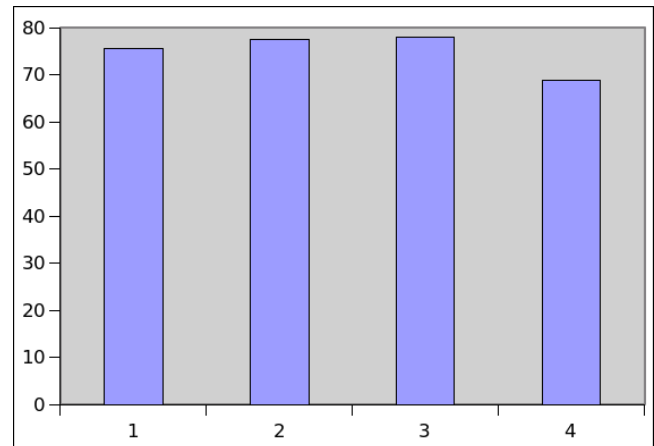


Figure 2. Average win rate

We also should run a least 1024 tournaments in order to the measurements be statistically relevant, reducing the standard deviation, keeping in mind that 10 tournaments consumed 01h53min59s or real time.