

Evaluating Heuristics

Summary

Three heuristic functions have been compared by using the supplied `tournament.py` script. I chose to base my heuristic functions on the number of blank spaces around the player. This seems like a valid heuristic as two out of three variants beat `ID_Improved`.

A note on `tournament.py`

The implementation of the tournament script is random-based, meaning the results for each simulation is not reproducible.

Detailed results for `ID_Improved`

Formula: `Moves - Opponent moves`

```
*****
Evaluating: ID_Improved
*****

Playing Matches:
-----
Match 1: ID_Improved vs Random      Result: 63 to 17
Match 2: ID_Improved vs MM_Null     Result: 48 to 32
Match 3: ID_Improved vs MM_Open     Result: 47 to 33
Match 4: ID_Improved vs MM_Improved Result: 36 to 44
Match 5: ID_Improved vs AB_Null     Result: 50 to 30
Match 6: ID_Improved vs AB_Open     Result: 43 to 37
Match 7: ID_Improved vs AB_Improved Result: 52 to 28

Results:
-----
ID_Improved      60.54%
```

Note that `ID_Improved` seems to be inferior to `MM_Improved`. This indicates that, at least on my hardware, the Iterative Deepening approach of `ID_Improved` does not give the expected speed up compared to a naive 3 level minimax approach. But looking at `ID_Improved` vs `AB_Improved` we do see an improvement in using Iterative Deepening compared to a level 5 alphabeta search. This discrepancy could also indicate that, given the `Improved` heuristic function, we do not see any particular gains for going deeper into the game state. This discrepancy warrants further investigation.

Given that `ID_Improved` beats all the `AB_*` agents, I see an indication that the iterative deepening approach of alphabeta search is an improvement over regular alphabeta search.

Detailed results for **Blank**

Formula: `Blank spaces`

Given the restrictive move for how the players move in this Isolation variant, I suspected that the number of blank spaces surrounding a player could be an indicator of how good a move it is, as the neighbouring cells for a given move would be available in moves ahead. I sought inspiration from this [https://en.wikipedia.org/wiki/Knight_\(chess\)](https://en.wikipedia.org/wiki/Knight_(chess)), but assigning the value 1 to each field.

```
*****
Evaluating: Blank
*****

Playing Matches:
-----
Match 1:  Blank    vs  Random      Result: 67 to 13
Match 2:  Blank    vs  MM_Null     Result: 57 to 23
Match 3:  Blank    vs  MM_Open     Result: 41 to 39
Match 4:  Blank    vs  MM_Improved Result: 36 to 44
Match 5:  Blank    vs  AB_Null     Result: 53 to 27
Match 6:  Blank    vs  AB_Open     Result: 53 to 27
Match 7:  Blank    vs  AB_Improved Result: 43 to 37

Results:
-----
Blank                62.50%
```

Again, we see that the `MM_Improved` proves quite the challenge. But worthy to note, is that the `Blank` heuristic is strong against the `Open` heuristic, indicating that I might be right in my hunch about using number of blank space as opposed to number of moves.

Detailed results for **Blank IMP**

Formula: `Blank spaces - Opponent blank spaces`

The naive approach performed pretty well, but lets try the same tactic as in the `Improved` heuristic and subtract the opponent's blank spaces. This means, that in a situation where the number of blank spaces is equal between moves, it would break the tie by limiting the number of blank spaces available to the opponent.

```
*****
Evaluating: Blank IMP
*****

Playing Matches:
-----
Match 1:  Blank IMP vs  Random      Result: 66 to 14
Match 2:  Blank IMP vs  MM_Null     Result: 59 to 21
```

```

Match 3: Blank IMP vs MM_Open Result: 49 to 31
Match 4: Blank IMP vs MM_Improved Result: 42 to 38
Match 5: Blank IMP vs AB_Null Result: 56 to 24
Match 6: Blank IMP vs AB_Open Result: 46 to 34
Match 7: Blank IMP vs AB_Improved Result: 51 to 29

```

```

Results:
-----

```

```

Blank IMP          65.89%

```

Overall, this agent did better than the `Blank` agent. Not by much, but enough to indicate that breaking the ties by limiting the opponent's blank spaces is a good idea.

Detailed results for `Blank MOV`

Formula: `Blank spaces / Opponent moves`

This heuristic builds upon the previous heuristic. Now we only break the ties if we actively restrict the opponent's available moves. But not only that, the division ensures that if we with a move can block the one of the opponent's legal moves, we will prioritise that move, even if it means fewer blank spaces.

```

*****
Evaluating: Blank MOV
*****

Playing Matches:
-----
Match 1: Blank MOV vs Random Result: 70 to 10
Match 2: Blank MOV vs MM_Null Result: 61 to 19
Match 3: Blank MOV vs MM_Open Result: 47 to 33
Match 4: Blank MOV vs MM_Improved Result: 41 to 39
Match 5: Blank MOV vs AB_Null Result: 52 to 28
Match 6: Blank MOV vs AB_Open Result: 53 to 27
Match 7: Blank MOV vs AB_Improved Result: 51 to 29

Results:
-----
Blank MOV          66.96%

```

Overall, this agent did better than the `Blank` and `Blank IMP` agents but not by much.

Evaluating `Blank MOV` against `ID_Improved`

I've evaluated all three heuristics against the `ID_Improved` heuristic.

```

Match 1: Blank vs ID_Improved Result: 44 to 36
Match 2: Blank IMP vs ID_Improved Result: 43 to 37

```

Interestingly, it seems that `Blank` and `Blank IMP` are relatively close in performance against `ID_Improved` and `Blank MOV` loses even though `Blank MOV` beat the baseline agents.

Depth of game tree

I've collected data for how deep a tree gets build on average. This metric is interesting because it gives an indication of how expensive a heuristic is to calculate. The lower execution time, more nodes we can visit.

Heuristic	Average	Max
ID_IMPROVED	89.98	1845
BLANK	85.10	1819
BLANK IMP	99.10	1831
BLANK MOV	88.95	1835

It seems all heuristics are somewhat similar in depth. Interestingly, `Blank IMP` goes deeper on average than `Blank`. Intuition says it should be the other way around. However, the results are consistent between multiple runs, so further investigation is warranted.

Conclusion

Based on the above discussions, I'd recommend going with the `Blank IMP` heuristic because:

1. It has the 2nd best performance against the baseline agents and `ID_Improved` of the three heuristics, but is very close to the best performance in both cases.
2. It considers both players available blank spaces.
3. It reaches the highest average depth, hopefully discovering win-conditions earlier than the opponent.

Future work

- Investigate the values of each field surrounding the agent when evaluating the blank spaces
- Better tournament evaluation, for reproducible comparisons.
- Investigate the discrepancy in performance against `ID_Improved` and against the baseline agents for `Blank MOV`.
- Investigate the performance of `Blank IMP` vs `Blank` for tree depth.