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WID3009

Heuristic Analysis Activity

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SYNOPSIS

The project aims at developing an adversarial search agent to play the game "Isolation". This project report focusses on the heuristics to be used in A* Search for minimax and alphabeta pruning.

Isolation is a deterministic, two-player game of perfect information in which the players alternate turns moving a single piece from one cell to another on a board. Whenever either player occupies a cell, that cell becomes blocked for the remainder of the game. The first player with no remaining legal moves loses, and the opponent is declared the winner.

This project uses a version of Isolation where each agent is restricted to L-shaped movements (like a knight in chess) on a rectangular grid (like a chess or checkerboard). The agents can move to any open cell on the board that is 2-rows and 1-column or 2-columns and 1-row away from their current position on the board. Movements are blocked at the edges of the board (the board does not wrap around), however, the player can "jump" blocked or occupied spaces (just like a knight in chess).

Additionally, agents will have a fixed time limit each turn to search for the best move and respond. If the time limit expires during a player's turn, that player forfeits the match, and the opponent wins. These rules are implemented in the isolation. Board class provided in the repository.

CUSTOM HEURISTICS

1. CUSTOM HEURISTIC

The heuristic based on the logic that player should have more moves in comparison to opponent and opponent should have less moves in comparison to player. The depth of the tree also taken into consideration.

It can be mathematically expressed as:

$$\alpha \frac{\text{len}(\text{my available moves})}{\text{len}(\text{available opponent moves})} - \frac{\text{len}(\text{available opponent moves})}{\text{len}(\text{my available moves})}, \text{ where } \alpha \in (1, \infty)$$

Maximizing above equation is equivalent to maximizing:

$$\beta [\text{len}(\text{my available moves})]^2 - [\text{len}(\text{my available moves})]^2 - \text{depth}(0.01), \text{ where } \beta \in (1, \infty)$$

The latter form has been implemented in the code with β chosen as 1.5 empirically.

EVALUATING HEURISTICS

The tournament.py script is used to evaluate the effectiveness of heuristic. The script measures relative performance of player in a round-robin tournament against several other pre-defined agents.

The performance of time-limited iterative deepening search is hardware dependent (faster hardware is expected to search deeper than slower hardware in the same amount of time). The script controls for these effects by also measuring the baseline performance of an agent called "ID_Improved" that uses Iterative Deepening and the improved_score heuristic from sample_players.py.

The tournament opponents are listed below:

- Random: An agent that randomly chooses a move each turn.
- MM_Null: CustomPlayer agent using fixed-depth minimax search and the null_score heuristic
- MM_Open: CustomPlayer agent using fixed-depth minimax search and the open_move_score heuristic
- MM_Improved: CustomPlayer agent using fixed-depth minimax search and the improved_score heuristic
- AB_Null: CustomPlayer agent using fixed-depth alpha-beta search and the null_score heuristic
- AB_Open: CustomPlayer agent using fixed-depth alpha-beta search and the open_move_score heuristic
- AB_Improved: CustomPlayer agent using fixed-depth alpha-beta search and the improved_score heuristic
- ID_Improved: CustomPlayer agent using iterative alpha-beta search and the improved_score heuristic
- Student1: CustomPlayer agent using iterative alpha-beta search and the [heuristic 1](#)
- Student2: CustomPlayer agent using iterative alpha-beta search and the [heuristic 2](#)
- Student3: CustomPlayer agent using iterative alpha-beta search and the [heuristic 3](#)
- Student4: CustomPlayer agent using iterative alpha-beta search and the [heuristic 4](#)
- Student5: CustomPlayer agent using iterative alpha-beta search and the [heuristic 5](#)
- Student6: CustomPlayer agent using iterative alpha-beta search and the [heuristic 6](#)
- Student7: CustomPlayer agent using iterative alpha-beta search and the [heuristic 7](#)
- MyHeuristic: CustomPlayer agent using iterative alpha-beta search and the [custom heuristic](#)

Since, running take a few hours to completed, the number of matches were reduced from 5 to 10.

RESULTS

The performance of various agents is as follow:

Agent	Performance	Rank
ID_Improved	59.29%	11
Student1	54.29%	12
Student2	59.29%	10
Student3	65.00%	5
Student4	67.86%	2
Student5	60.00%	9
Student6	63.57%	7
Student7	62.86%	8
MyHeuristic	65.71%	3
TeamMember1 (Syafik)	65.00%	4
TeamMember2 (Hakim)	64.29%	6
TeamMember3 (Aqiff)	67.86%	1

All the custom heuristics perform better than ID_Improved by a reasonable margin as can be seen in the above table. Student4 and my team member Aqiff performs better than other including my custom heuristic.

APPENDICES

A. APPENDIX: EVALUATION RESULT

This script evaluates the performance of the custom heuristic function by comparing the strength of an agent using iterative deepening (ID) search with alpha-beta pruning against the strength rating of agents using other heuristic functions. The `ID_Improved` agent provides a baseline by measuring the performance of a basic agent using Iterative Deepening and the "improved" heuristic (from lecture) on your hardware. The `Student` agent then measures the performance of Iterative Deepening and the custom heuristic against the same opponents.

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

Playing Matches:

```
-----
Match 1: ID_Improved vs Random      Result: 16 to 4
Match 2: ID_Improved vs MM_Null     Result: 14 to 6
Match 3: ID_Improved vs MM_Open     Result: 9 to 11
Match 4: ID_Improved vs MM_Improved Result: 7 to 13
Match 5: ID_Improved vs AB_Null     Result: 12 to 8
Match 6: ID_Improved vs AB_Open     Result: 12 to 8
Match 7: ID_Improved vs AB_Improved Result: 13 to 7
```

Results:

```
-----
ID_Improved          59.29%
```

```
*****
Evaluating: Student1
*****
```

Playing Matches:

```
-----
Match 1: Student1 vs Random      Result: 15 to 5
Match 2: Student1 vs MM_Null     Result: 12 to 8
Match 3: Student1 vs MM_Open     Result: 8 to 12
Match 4: Student1 vs MM_Improved Result: 7 to 13
Match 5: Student1 vs AB_Null     Result: 13 to 7
Match 6: Student1 vs AB_Open     Result: 11 to 9
Match 7: Student1 vs AB_Improved Result: 10 to 10
```

Results:

```
-----
Student1             54.29%
```

```
*****
Evaluating: Student2
*****
```

Playing Matches:

```
-----
Match 1: Student2 vs Random Result: 17 to 3
Match 2: Student2 vs MM_Null Result: 13 to 7
Match 3: Student2 vs MM_Open Result: 10 to 10
Match 4: Student2 vs MM_Improved Result: 9 to 11
Match 5: Student2 vs AB_Null Result: 14 to 6
Match 6: Student2 vs AB_Open Result: 10 to 10
Match 7: Student2 vs AB_Improved Result: 10 to 10
```

Results:

```
-----
Student2 59.29%
```

```
*****
Evaluating: Student3
*****
```

Playing Matches:

```
-----
Match 1: Student3 vs Random Result: 18 to 2
Match 2: Student3 vs MM_Null Result: 14 to 6
Match 3: Student3 vs MM_Open Result: 12 to 8
Match 4: Student3 vs MM_Improved Result: 9 to 11
Match 5: Student3 vs AB_Null Result: 12 to 8
Match 6: Student3 vs AB_Open Result: 14 to 6
Match 7: Student3 vs AB_Improved Result: 12 to 8
```

Results:

```
-----
Student3 65.00%
```

```
*****
Evaluating: Student4
*****
```

Playing Matches:

```
-----
Match 1: Student4 vs Random Result: 20 to 0
Match 2: Student4 vs MM_Null Result: 12 to 8
Match 3: Student4 vs MM_Open Result: 14 to 6
Match 4: Student4 vs MM_Improved Result: 10 to 10
Match 5: Student4 vs AB_Null Result: 11 to 9
Match 6: Student4 vs AB_Open Result: 16 to 4
Match 7: Student4 vs AB_Improved Result: 12 to 8
```


Results:

Student4 67.86%

Evaluating: Student5

Playing Matches:

Match 1:	Student5	vs	Random	Result: 15 to 5
Match 2:	Student5	vs	MM_Null	Result: 11 to 9
Match 3:	Student5	vs	MM_Open	Result: 13 to 7
Match 4:	Student5	vs	MM_Improved	Result: 10 to 10
Match 5:	Student5	vs	AB_Null	Result: 13 to 7
Match 6:	Student5	vs	AB_Open	Result: 12 to 8
Match 7:	Student5	vs	AB_Improved	Result: 10 to 10

Results:

Student5 60.00%

Evaluating: Student6

Playing Matches:

Match 1:	Student6	vs	Random	Result: 16 to 4
Match 2:	Student6	vs	MM_Null	Result: 15 to 5
Match 3:	Student6	vs	MM_Open	Result: 10 to 10
Match 4:	Student6	vs	MM_Improved	Result: 11 to 9
Match 5:	Student6	vs	AB_Null	Result: 14 to 6
Match 6:	Student6	vs	AB_Open	Result: 14 to 6
Match 7:	Student6	vs	AB_Improved	Result: 9 to 11

Results:

Student6 63.57%

Evaluating: Student7

Playing Matches:

Match 1:	Student7	vs	Random	Result: 18 to 2
Match 2:	Student7	vs	MM_Null	Result: 14 to 6
Match 3:	Student7	vs	MM_Open	Result: 7 to 13
Match 4:	Student7	vs	MM_Improved	Result: 11 to 9
Match 5:	Student7	vs	AB_Null	Result: 13 to 7
Match 6:	Student7	vs	AB_Open	Result: 11 to 9

Match 7: Student7 vs AB_Improved Result: 14 to 6

Results:

Student7 62.86%

Evaluating: MyHeuristic

Playing Matches:

Match 1: MyHeuristic vs Random Result: 15 to 5
Match 2: MyHeuristic vs MM_Null Result: 15 to 5
Match 3: MyHeuristic vs MM_Open Result: 11 to 9
Match 4: MyHeuristic vs MM_Improved Result: 9 to 11
Match 5: MyHeuristic vs AB_Null Result: 14 to 6
Match 6: MyHeuristic vs AB_Open Result: 13 to 7
Match 7: MyHeuristic vs AB_Improved Result: 15 to 5

Results:

MyHeuristic 65.71%

Evaluating: TeamMember1

Playing Matches:

Match 1: TeamMember1 vs Random Result: 16 to 4
Match 2: TeamMember1 vs MM_Null Result: 14 to 6
Match 3: TeamMember1 vs MM_Open Result: 14 to 6
Match 4: TeamMember1 vs MM_Improved Result: 11 to 9
Match 5: TeamMember1 vs AB_Null Result: 12 to 8
Match 6: TeamMember1 vs AB_Open Result: 12 to 8
Match 7: TeamMember1 vs AB_Improved Result: 12 to 8

Results:

TeamMember1 65.00%

Evaluating: TeamMember2

Playing Matches:

Match 1: TeamMember2 vs Random Result: 14 to 6
Match 2: TeamMember2 vs MM_Null Result: 16 to 4

Match 3:	TeamMember2	vs	MM_Open	Result: 12 to 8
Match 4:	TeamMember2	vs	MM_Improved	Result: 8 to 12
Match 5:	TeamMember2	vs	AB_Null	Result: 14 to 6
Match 6:	TeamMember2	vs	AB_Open	Result: 15 to 5
Match 7:	TeamMember2	vs	AB_Improved	Result: 11 to 9

Results:

TeamMember2 64.29%

Evaluating: TeamMember3

Playing Matches:

Match 1:	TeamMember3	vs	Random	Result: 18 to 2
Match 2:	TeamMember3	vs	MM_Null	Result: 15 to 5
Match 3:	TeamMember3	vs	MM_Open	Result: 10 to 10
Match 4:	TeamMember3	vs	MM_Improved	Result: 9 to 11
Match 5:	TeamMember3	vs	AB_Null	Result: 18 to 2
Match 6:	TeamMember3	vs	AB_Open	Result: 12 to 8
Match 7:	TeamMember3	vs	AB_Improved	Result: 13 to 7

Results:

TeamMember3 67.86%