# **Heuristic Analysis**

There are 4 heuristics defined in this project:

- Minimize\_opponent\_hits
- Simplest\_aggressive
- Maximize\_winning
- Minimize\_losing
- Weighted\_combination

# 1. Minimize\_opponent\_hits: Student0

I tried something new with this one. Instead of just subtracting the number of opponent moves from my moves and then minimizing the whole thing, I first find the number of blank spaces left on board (NBS)

Then I subtract my moves from NBS to get all the blank spaces which exclude my moves. I am trying to maximize those blank spaces which actually aren't contributing for opponent moves.

I am penalizing more and more on the basis of how much fraction of not\_my\_moves variable is actually contributing to opponent's move. Contribution is directly proportional to the penalty. There's actually a lot that you can tweak in this heuristic. For e.g. you can add a reward parameter for main player apart from penalty for opponent to make the heuristic more aggressive.

**2. Simplest\_aggressive:** By this heuristic, we are trying to minimize opponent's moves. (Described in lectures) Instead of subtracting the no. of opponent moves (OM) from the main player's moves (PM) we're using a more aggressive approach. Before subtracting OM from PM, we multiply it by 2. This way, we're rewarding the states which have less and less number of OM left which in turn makes it easier for main player to win. It can be mathematically expressed as:

len(main\_player\_moves) - alpha \* len(opponent's moves)

As described above, the value of alpha is chosen as 2.

**3. Maximize\_winning:** By this approach, we're trying to maximize the ratio of main player's moves to opponent's moves.

It can be mathematically expressed as:

len(main\_player\_moves) / len(opponent's moves)

- **4. Minimize\_losing:** Here, we're trying to minimize the ratio of opponent moves to main player's moves.
  - len(opponent's moves) / len(main\_player\_moves)

# 5. Weighted\_combination:

This can be expressed as:

# alpha \* maximize\_winning - beta \* minimize\_losing

As we can see that this is much effective heuristic, since it combines both of the above heuristics.

## **Results:**

Agent	Average Performance
ID_Improved	67.50+71.43+67.14 = 68.69%
Student0	69.64+71.43+74.29 = 71.78%
Student1	68.21+67.86+68.21 = 68.09%
Student2	67.50+69.64+65 = 67.38%
Student3	67.14+65+67.86 = 66.66%
Student4	67.50+76.43+72.86 = 72.26%

## **Analysis**

## 1. Computational complexity order:

Weighted\_combination > Minimize\_opponent\_hits > Minimize\_losing = Maximize\_winning > Simplest\_aggressive

Minimize\_losing and Maximize\_winning are computationally equivalent. We can see that Simplest\_aggressive heuristic beats both Minimize\_losing and Maximize\_winning. Probably because it's computationally faster than both. As explained in lectures, the factor that 2 is being multiplied by opponent's moves before subtraction from my\_moves, makes it more aggressive. As in, it tries to reach a final state way faster than other 2 mentioned heuristics.

As we can see, for three attempts, average performances of Student4 and Student0 are quite well. Student4 beats all other heuristics, that's why I have submitted it. As I explained in Minimize\_opponent\_hits heuristic that adding an extra reward parameter apart from penalty would actually make the heuristic more powerful. This will also help it to reach the goal state faster.

In **weighted\_combination** heuristic, both heuristics, **maximize\_winning** and **minimize\_losing** are working together in favor of main player which makes it most powerful. It also helps to reach a final state faster.

# Result of tournament.py:

## Attempt 1

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.

#### Playing Matches:

Result: 33 to 7 Match 1: ID Improved vs Random Match 2: ID Improved vs MM Null Result: 27 to 13 Result: 27 to 13 Match 3: ID Improved vs MM Open Match 4: ID Improved vs MM Improved Result: 26 to 14 Result: 26 to 14 Match 5: ID Improved vs AB Null Match 6: ID Improved vs AB Open Result: 27 to 13 Match 7: ID Improved vs AB Improved Result: 23 to 17

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Results:
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------ID Improved 67.50%

## Playing Matches:

N	Match	1:	Student0	vs	Random	Result:	34	to	6
N	Match	2:	Student0	vs	MM_Null	Result:	33	to	7
ľ	Match	3:	Student0	vs	MM_Open	Result:	26	to	14
ľ	Match	4:	Student0	vs	MM_Improved	Result:	25	to	15
N	Match	5:	Student0	vs	AB_Null	Result:	28	to	12
N	Match	<b>6:</b>	Student0	vs	AB_Open	Result:	23	to	17
ľ	Match	7:	Student0	vs	AB Improved	Result:	26	to	14

### Results:

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Student0 69.64%

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## Playing Matches:

Match	1:	Student1	vs	Random	Result:	34	to	6
Match	2:	Student1	vs	MM_Null	Result:	30	to	10
Match	3:	Student1	vs	MM_Open	Result:	25	to	15
Match	4:	Student1	vs	MM_Improved	Result:	24	to	16
Match	5:	Student1	vs	AB_Null	Result:	26	to	14
Match	6 <b>:</b>	Student1	vs	AB_Open	Result:	25	to	15
Match	7:	Student1	vs	AB_Improved	Result:	27	to	13

## Results:

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Student1 68.21%

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# Playing Matches:

Match 1:	Student2	vs	Random	Result:	33 to 7
Match 2:	Student2	vs	MM_Null	Result:	29 to 11
Match 3:	Student2	vs	MM_Open	Result:	20 to 20
Match 4:	Student2	vs	MM_Improved	Result:	27 to 13
Match 5:	Student2	vs	AB_Null	Result:	35 to 5
Match 6:	Student2	vs	AB_Open	Result:	21 to 19
Match 7:	Student2	vs	AB Improved	Result:	24 to 16

## Results:

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Student2 67.50%

## Playing Matches:

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Match	1:	Student3	vs	Random	Result:	30	to	10
Match	2:	Student3	vs	MM_Null	Result:	30	to	10
Match	3:	Student3	vs	MM_Open	Result:	24	to	16
Match	4:	Student3	vs	MM_Improved	Result:	27	to	13
Match	5:	Student3	vs	AB_Null	Result:	29	to	11
Match	<b>6:</b>	Student3	vs	AB_Open	Result:	25	to	15
Match	7:	Student3	vs	AB Improved	Result:	23	to	17

## Results:

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Student3 67.14%

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## Playing Matches:

Match	1:	Student4	vs	Random	Result:	32	to	8
Match	2:	Student4	vs	MM_Null	Result:	32	to	8
Match	3:	Student4	vs	MM_Open	Result:	28	to	12
Match	4:	Student4	vs	MM_Improved	Result:	25	to	15
Match	5:	Student4	vs	AB_Null	Result:	29	to	11
Match	<b>6:</b>	Student4	vs	AB_Open	Result:	21	to	19
Match	7:	Student4	vs	AB_Improved	Result:	22	to	18

#### Results:

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Student4 67.50%

## Attempt 2

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.

## Playing Matches:

\_\_\_\_\_ Match 1: ID Improved vs Random Result: 35 to 5 Result: 29 to 11 Match 2: ID Improved vs MM Null Result: 25 to 15 Match 3: ID Improved vs MM Open Result: 26 to 14 Match 4: ID Improved vs MM Improved Match 5: ID Improved vs AB Null Result: 33 to 7 AB Open Result: 26 to 14 Match 6: ID Improved vs Match 7: ID Improved vs AB Improved Result: 26 to 14

#### Results:

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ID Improved 71.43%

## Playing Matches:

\_\_\_\_\_ Match 1: Student0 Result: 34 to 6 VS Random Match 2: Student0 MM Null Result: 36 to 4 vs Result: 28 to 12 Match 3: Student0 vs MM Open Result: 23 to 17 Match 4: Student0 vs MM Improved Result: 32 to 8 Match 5: Student0 vs AB Null Match 6: Student0 vs AB Open Result: 23 to 17 Match 7: Student0 Result: 24 to 16 vs AB Improved

#### Results:

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Student0 71.43%

## Playing Matches:

-----Match 1: Student1 vs Random Result: 35 to 5 Match 2: MM Null Result: 31 to 9 Student1 vs Match 3: Result: 24 to 16 Student1 MM Open vs Result: 25 to 15 Match 4: vs MM\_Improved Student1 Result: 27 to 13 Match 5: Student1 AB Null vs Result: 25 to 15 Match 6: Student1 vs AB Open Match 7: vs AB\_Improved Result: 23 to 17 Student1

### Results:

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Student1 67.86%

### Playing Matches:

Match 1: Studen

Result: 31 to 9 Student2 vs Random Result: 35 to 5 Match 2: Student2 vs MM Null Match 3: Student2 MM Open Result: 27 to 13 vs Match 4: Student2 Result: 23 to 17 vs MM Improved Match 5: Student2 AB Null Result: 30 to 10 vs Result: 25 to 15 Match 6: Student2 AB Open vs Match 7: Student2 vs AB Improved Result: 24 to 16

### Results:

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Student2 69.64%

## Playing Matches:

Match 1: Result: 31 to 9 Student3 vs Random MM Null Result: 28 to 12 Match 2: Student3 vs Match 3: Student3 vs MM Open Result: 25 to 15 vs MM Improved Result: 24 to 16 Match 4: Student3 Match 5: AB Null Result: 27 to 13 Student3 vs Match 6: AB Open Result: 23 to 17 Student3 vs

Match 7: Student3 vs AB Improved Result: 24 to 16

### Results:

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Student3 65.00%

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## Playing Matches:

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Match 1:	Student4	vs	Random	Result:	36	to	4
Match 2:	Student4	vs	MM_Null	Result:	36	to	4
Match 3:	Student4	vs	MM_Open	Result:	30	to	10
Match 4:	Student4	vs	MM_Improved	Result:	30	to	10
Match 5:	Student4	vs	AB_Null	Result:	37	to	3
Match 6:	Student4	vs	AB_Open	Result:	22	to	18
Match 7:	Student4	vs	AB_Improved	Result:	23	to	17

#### Results:

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Student4 76.43%

#### Attempt 3

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.

Playing Matches:

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Match 1: ID\_Improved vs Random Result: 35 to 5

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Match 2: ID Improved vs
                          MM Null
                                          Result: 29 to 11
Match 3: ID Improved vs
                          MM Open
                                          Result: 28 to 12
Match 4: ID Improved vs MM Improved
                                          Result: 24 to 16
Match 5: ID Improved vs
                          AB Null
                                          Result: 26 to 14
Match 6: ID Improved vs
                                          Result: 24 to 16
                          AB Open
Match 7: ID Improved vs AB Improved
                                          Result: 22 to 18
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### Results:

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ID Improved 67.14%

## Playing Matches:

\_\_\_\_\_ Match 1: Student0 VS Random Result: 37 to 3 Match 2: Student0 Result: 34 to 6 vs MM Null Match 3: Student0 vs MM Open Result: 28 to 12 Result: 27 to 13 Match 4: Student0 vs MM Improved Match 5: Student0 AB Null Result: 26 to 14 vs AB Open Result: 29 to 11 Match 6: Student0 vs Match 7: vs AB Improved Result: 27 to 13 Student0

#### Results:

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Student0 74.29%

### Playing Matches:

-----Match 1: Student1 Result: 34 to 6 VS Random Match 2: Student1 Result: 31 to 9 vs MM Null Match 3: Student1 Result: 26 to 14 VS MM Open Match 4: Student1 vs MM Improved Result: 24 to 16 Match 5: Student1 Result: 28 to 12 vs AB Null Match 6: Student1 AB Open Result: 24 to 16 vs Match 7: Student1 Result: 24 to 16 vs AB Improved

#### Results:

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Student1 68.21%

Result: 34 to 6

Result: 28 to 12

Result: 27 to 13

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*******
 Evaluating: Student2
*******
Playing Matches:
 Match 1:
                                      Result: 31 to 9
         Student2
                    vs
                         Random
 Match 2: Student2
                         MM Null
                                      Result: 34 to 6
                    vs
 Match 3: Student2
                                      Result: 25 to 15
                         MM Open
                    VS
 Match 4: Student2
                    vs MM Improved
                                      Result: 21 to 19
 Match 5: Student2
                         AB Null
                                      Result: 25 to 15
                    vs
 Match 6: Student2
                    vs
                         AB Open
                                      Result: 22 to 18
 Match 7: Student2
                                      Result: 24 to 16
                    vs AB Improved
Results:
_____
Student2
                  65.00%
*******
 Evaluating: Student3
*******
Playing Matches:
_____
 Match 1: Student3
                                      Result: 35 to 5
                    vs
                         Random
 Match 2: Student3
                                      Result: 28 to 12
                         MM Null
                    VS
 Match 3: Student3
                         MM Open
                                      Result: 24 to 16
                    vs
                    vs MM Improved
                                      Result: 25 to 15
 Match 4: Student3
 Match 5: Student3
                    vs
                         AB Null
                                      Result: 34 to 6
 Match 6: Student3
                         AB Open
                                      Result: 24 to 16
                    vs
 Match 7: Student3
                    vs AB Improved
                                      Result: 20 to 20
Results:
_____
Student3
                  67.86%
********
 Evaluating: Student4
*******
Playing Matches:
_____
 Match 1: Student4
                    vs
                         Random
                                      Result: 36 to 4
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Match 2: Student4

Student4

Student4

Match 3:

Match 4:

vs

VS

MM Null

MM Open

vs MM Improved

# Simmi Mourya, Udacity AIND Isolation

Match 5:	Student4	vs	AB_Null	Result:	30	to	10
Match 6:	Student4	vs	AB_Open	Result:	26	to	14
Match 7:	Student4	vs .	AB Improved	Result:	23	to	17

Results:

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Student4 72.86%