Heuristics Analysis:

Totally, I have written 3 heuristics, and all three heuristics have below 2 parts in common:

- 1. If the Non-Opposite player is the winner, return an infinite reward (+inf)
- 2. If the Opposite player is the winner, return an infinite penalty (-inf)

However, if the game hasn't been ended yet, the 3 heuristics differ as follows:

- 1. The first heuristic, custom_score_1, assigns a relatively High Reward for the number of moves the player has, and low penalty for the number of moves opponent has.
- 2. The second heuristic, custom_score_2, assigns a relatively low Reward for the number of moves the player has, and High penalty for the number of moves opponent has.
- 3. The third heuristic, custom_score_3, assigns Equal Reward for the number of moves the player has, and for the number of moves opponent has.

The reward and penalties are scaled according the variables reward_cofficient and penalty_coefficient respectively.

Out of the 3 heuristics above, I chose the second heuristic custom_score_2, where in reward_cofficient is 2 and penalty coefficient is 4.

The numbers 2 and 4 were arrived by empirically verifying the results, assuming to be a sort of a hyperparameter that I have to tune.

The time complexity of all the 3 heuristics remain the same, as they are almost identical, but only differ by the coefficients.

The screenshots of the results of best heuristic is below:

```
Select Command Prompt
neuristic (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
                                                      Result: 0 to 20
                                     Random
 Match 2: ID_Improved vs MM_Null
Match 3: ID_Improved vs MM_Open
Match 4: ID_Improved vs MM_Improved
                                                      Result: 0 to 20
                                                      Result: 0 to 20
                                                      Result: 0 to 20
  Match 5: ID_Improved vs AB_Null
                                                      Result: 0 to 20
  Match 6: ID_Improved vs AB_Open
Match 7: ID_Improved vs AB_Improved
                                                      Result: 4 to 16
Result: 2 to 18
Results:
ID_Improved
                            4.29%
  Evaluating: Student
Playing Matches:
  Match 1:
                 Student
                                     Random
                                                      Result: 0 to 20
 Match 2: Student vs MM_Null Result: 0 to 20 Match 3: Student vs MM_Open Result: 0 to 20
                                   MM Null
ournament.py:100: UserWarning: One or more agents lost a match this round due to timeout. The get function must return before time_left() reaches 0 ms. You will need to leave some time for the forto return, and may need to increase this margin to avoid timeouts during tournament play.
  warnings.warn(TIMEOUT_WARNING)
 Match 4: Student vs MM_Improved
Match 5: Student vs AB_Null
Match 6: Student vs AB_Open
                                                      Result: 0 to 20
                                                     Result: 1 to 19
                                                      Result: 5 to 15
  Match 7:
                 Student vs AB Improved
                                                      Result: 6 to 14
Results:
```

8.57%

datascience_3.5) C:\Users\kabangal\Desktop\AIND\AIND\P2>

Student