report

June 9, 2020

1 Build a Game Playing Agent - Knights Isolation

1.1 James McGuigan

• Source: https://github.com/JamesMcGuigan/udacity-artificial-intelligence/blob/master/Projects/3 Adversarial%20Search/

2 Unit Tests

```
[1]: from agents.AlphaBetaPlayer import AlphaBetaAreaPlayer, AlphaBetaPlayer,

MinimaxPlayer
from agents.DistancePlayer import DistancePlayer, GreedyDistancePlayer
from agents.MCTS import MCTSMaximum, MCTSMaximumHeuristic, MCTSRandom,

MCTSRandomHeuristic
from agents.UCT import UCTPlayer
from sample_players import GreedyPlayer, RandomPlayer
from run_backpropagation import run_backpropagation, TEST_AGENTS
from isolation import Agent, logger
from run_match_sync import play_sync
import time
import gc

%load_ext autoreload
%autoreload 2
```

```
[2]: ! python3 -m unittest -v
```

test_get_action_midgame (tests.test_my_custom_player.CustomPlayerGetActionTest) get_action() calls self.queue.put() before timeout in a game in progress ... ok test_get_action_player1 (tests.test_my_custom_player.CustomPlayerGetActionTest) get_action() calls self.queue.put() before timeout on an empty board ... ok test_get_action_player2 (tests.test_my_custom_player.CustomPlayerGetActionTest) get_action() calls self.queue.put() before timeout as player 2 ... ok test_get_action_terminal (tests.test_my_custom_player.CustomPlayerGetActionTest) get_action() calls self.queue.put() before timeout when the game is over ... ok test_custom_player (tests.test_my_custom_player.CustomPlayerPlayTest)

CustomPlayer successfully completes a game against itself ... ok

-----Ran 5 tests in 18.558s

3 Infrastructure

OK

- run_match_sync.py
- run_backpropagation.py
- agents/DataSavePlayer.py

I've rewritten run_match.py with a synchronous implementation in run_match_sync.py and run_backpropagation.py. using signals rather than multithreading.Pool(), which allows for better profiling and a 2x performance speedup.

I have improved the CLI flags and logging ouput, with an extra --verbose flag which can be used to ASCII print out the board state after each turn.

DataSavePlayer now handles loading and atexit autosaving of cls.data, whilst also gzipping contents

4 Advanced Heuristic

• agents/AlphaBetaPlayer.py

What features of the game does your heuristic incorporate, and why do you think those features matter in evaluating states during search?

The main analogy is with the game of Go. The goal is to surround your opponent and capture a larger territory.

Recursively computing liberties several moves ahead shows the area of the board that the opponent could potentially escape to.

At depth=1 this heuristic_area() is equivalent to #my_moves - #opponent_moves

Early in the game the opponent effectively has access to the entire board, making this heuristic ineffective, hence max_area is used in addition to depth to shortcircuit the computational cost of expanding breadth-first search to all possible future moves on a mostly empty board when neither side is trapped.

This heuristic is endgame focused. It solves the simplified subproblem of local search without an adversary, and provides an upper-bound estimate for the difference in maximum number of total moves each player has remaining. It reaches maximum value for moves that trap a player within a self-contained section whilst leaving the other a means of escape. The player in the smaller territory will run out of moves first.

The other major impact on the performance of this agent is the addition of alphabeta pruning with the aggressive use of caching. This avoids the computation expense of recomputing previously

explored subtrees, and by caching on @classmethod parts of the cache can even be reused between runs. The net effect of this is to increase the maximum depth of iterative deepening before the timeout to beyond what MinimaxPlayer(depth=3) can compute.

Improvements over previous submission: - Iterative Deepening checks terminates early if action score is infinite - Caching has been redone: now keyed on (player_id,state) and ignores depth/alpha/beta - Alphabeta caching only stores infinite values, which is effectively an endgame table - Caching is persisted to disk ./data/*.zi p.pickle via DataSavePlayer base class - Persisted caching means Alphabeta can pretrained with a higher timeout before the match

Performance improvements mean the algorithm now dominates the course Minimax implementation - First percentage is the total winrate, second percentage is the rolling average winrate

```
[3]: |rm -f ./data/A*.pickle
```

Greedy vs Minimax (at depth 2) depends on who gets the first turn. At standard depth 3, it has a 100% winrate.

```
[4]: | python3 ./run_backpropagation.py -a GREEDY -o MINIMAX --progress -r 70

------ match_id: 70 | 72s | 0% -> 0% | Greedy vs Minimax

AlphaBeta gets 86%+ winrate vs both Greedy and Minimax
```

[6]: | python3 ./run_backpropagation.py -a ALPHABETA -o MINIMAX --progress -r 70

Alpha Beta Area Player gets near 100% win
rate vs Greedy and Minimax, plus 70% vs Alpha Beta Player

368483

```
[8]: | python3 ./run backpropagation.py -a AREA -o MINIMAX --progress -r 70
   loaded: ./data/AlphaBetaAreaPlayer.zip.pickle
                                           | 5.9MB in 2.0s | entries:
   368483
   70 | 327s | 97% -> 96% | AlphaBeta Area vs Minimax
         ./data/AlphaBetaAreaPlayer.zip.pickle
                                           | 13.8MB in 18.1s | entries:
   873487
[9]: ! python3 ./run_backpropagation.py -a AREA -o ALPHABETA --progress -r 70
   loaded: ./data/AlphaBetaPlayer.zip.pickle
                                           | 1.6MB in 0.5s | entries:
   107152
   loaded: ./data/AlphaBetaAreaPlayer.zip.pickle
                                         | 13.8MB in 6.2s | entries:
   873487
   70 | 623s | 56% -> 58% | AlphaBeta Area vs AlphaBeta
         ./data/AlphaBetaPlayer.zip.pickle
                                       | 4.0MB in 4.5s | entries:
   wrote:
   261045
   wrote:
          ./data/AlphaBetaAreaPlayer.zip.pickle | 22.9MB in 25.2s | entries:
   1452080
```

5 Depth Analysis

- MiniMax course default is depth 3
- AlphaBeta
 - has trouble getting past depth 1 on the first turn
 - can get to depth 5 on the second turn
 - remains at depth 4-6 for the early game
 - grows to depth 6-9 during the mid-game
 - expands out to max depth of 14 before finding a -inf lose condition
- AlphaBetaArea
 - is usually depth-2 compared to AlphaBeta
 - the area heuristic effectively adds an extra 4 layers of hidden depth
 - so for the same CPU cost, AlphaBetaArea has a depth 2 advantage
 - finds the inf win condition before AlphaBeta and at a lower depth

203452 loaded: ./data/AlphaBetaAreaPlayer.zip.pickle | 18.0MB in 6.1s | entries:

```
AlphaBetaPlayer
                     | depth: 1
AlphaBetaAreaPlayer
                     | depth: 1 2
AlphaBetaPlayer
                     | depth: 1 2 3 4
AlphaBetaAreaPlayer
                       depth: 1 2 3 4
AlphaBetaPlayer
                       depth: 1 2 3 4 5
AlphaBetaAreaPlayer
                     | depth: 1 2 3 4 5
AlphaBetaPlayer
                       depth: 1 2 3 4 5
AlphaBetaAreaPlayer
                       depth: 1 2 3 4 5
                       depth: 1 2 3 4
AlphaBetaPlayer
AlphaBetaAreaPlayer
                       depth: 1 2 3
                       depth: 1 2 3 4 5
AlphaBetaPlayer
                       depth: 1 2 3
AlphaBetaAreaPlayer
AlphaBetaPlayer
                     | depth: 1 2 3 4
                       depth: 1 2 3
AlphaBetaAreaPlayer
                       depth: 1 2 3 4 5
AlphaBetaPlayer
AlphaBetaAreaPlayer
                     | depth: 1 2 3
                       depth: 1 2 3 4 5
AlphaBetaPlayer
                       depth: 1 2 3
AlphaBetaAreaPlayer
AlphaBetaPlayer
                       depth: 1 2 3 4 5
AlphaBetaAreaPlayer
                     | depth: 1 2 3
AlphaBetaPlayer
                       depth: 1 2 3 4 5
AlphaBetaAreaPlayer
                       depth: 1 2 3
                     | depth: 1 2 3 4 5
AlphaBetaPlayer
                       depth: 1 2 3
AlphaBetaAreaPlayer
                       depth: 1 2 3 4 5 6
AlphaBetaPlayer
AlphaBetaAreaPlayer
                       depth: 1 2 3 4
AlphaBetaPlayer
                     | depth: 1 2 3 4 5 6
                       depth: 1 2 3 4
AlphaBetaAreaPlayer
                       depth: 1 2 3 4 5 6
AlphaBetaPlayer
                     | depth: 1 2 3 4
AlphaBetaAreaPlayer
AlphaBetaPlayer
                       depth: 1 2 3 4 5 6 7
AlphaBetaAreaPlayer
                       depth: 1 2 3
                       depth: 1 2 3 4 5 6 7
AlphaBetaPlayer
AlphaBetaAreaPlayer
                     | depth: 1 2 3 4 5
AlphaBetaPlayer
                       depth: 1 2 3 4 5 6 7
                       depth: 1 2 3 4 5
AlphaBetaAreaPlayer
                     | depth: 1 2 3 4 5 6 7
AlphaBetaPlayer
AlphaBetaAreaPlayer
                     | depth: 1 2 3 4
                     | depth: 1 2 3 4 5 6 7 8
AlphaBetaPlayer
AlphaBetaAreaPlayer
                       depth: 1 2 3 4 5
                     | depth: 1 2 3 4 5 6 7 8
AlphaBetaPlayer
AlphaBetaAreaPlayer
                       depth: 1 2 3 4 5 6 7
                       depth: 1 2 3 4 5 6 7 8 9 10
AlphaBetaPlayer
AlphaBetaAreaPlayer
                       depth: 1 2 3 4 5 6 7 8 9
AlphaBetaPlayer
                       depth: 1 2 3 4 5 6 7 8 9 10 11 12
                     | depth: 1 2 3 4 5 6 7 8 9
AlphaBetaAreaPlayer
```

```
AlphaBetaPlayer
                     | depth: 1 2 3 4 5 6 7 8 9 10
                     | depth: 1 2 3 4 5 6 7 8 9
AlphaBetaAreaPlayer
AlphaBetaPlayer
                     | depth: 1 2 3 4 5 6 7 8 9 10 11 12
AlphaBetaAreaPlayer
                     | depth: 1 2 3 4 5 6 7 8
AlphaBetaPlayer
                     | depth: 1 2 3 4 5 6 7 8 9 10 11
AlphaBetaAreaPlayer
                     | depth: 1 2 3 4 5 6 7 8
AlphaBetaPlayer
                     | depth: 1 2 3 4 5 6 7 8 9 10 11 12 13 14
AlphaBetaAreaPlayer
                     | depth: 1 2 3 4 5 6 7 8 9 10 11
AlphaBetaPlayer
                     | depth: 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20
21 22 inf
                     | depth: 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17
AlphaBetaAreaPlayer
AlphaBetaPlayer
                     | depth: 1 inf
AlphaBetaAreaPlayer
                     | depth: 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19
-inf
AlphaBetaPlayer
                     | depth: 1 inf
AlphaBetaAreaPlayer
                     | depth: 1 -inf
AlphaBetaPlayer
                     | depth: 1 inf
AlphaBetaAreaPlayer | depth: 1 -inf
AlphaBetaPlayer
                     | depth: 1 inf
AlphaBetaAreaPlayer | depth: 1 -inf
AlphaBetaPlayer
                     | depth: 1 inf
AlphaBetaAreaPlayer | depth: 1 -inf
AlphaBetaPlayer
                     | depth: 1 inf
AlphaBetaAreaPlayer | depth: 1 -inf
AlphaBetaPlayer
                     | depth: 1 inf
AlphaBetaAreaPlayer
                     | depth: 1 -inf
AlphaBetaPlayer
                     | depth: 1 inf
AlphaBetaAreaPlayer
                     | depth: 1 -inf
```

If this can be done, then it turns out that alpha—beta needs to examine only O(bm/2) nodes to pick the best move, instead of O(bm) for minimax. This means that the effective branching factor becomes \sqrt{b} instead of b—for chess, about 6 instead of 35. Put another way, alpha—beta can solve a tree roughly twice as deep as minimax in the same amount of time. - Artificial Intelligence: A Modern Approach (p160)

In practice, Alpha Beta gets an smaller depth lead over MiniMax with an average of 1 and range of 0-2

```
[32]: # don't load existing caches

class AlphaBeta(AlphaBetaPlayer):

search_fn = 'alphabeta'
heuristic_fn = 'heuristic_liberties' # 'heuristic_liberties' /

→ 'heuristic_area'
verbose_depth = True
def load(cls): pass
def save(cls): pass

class MiniMax(AlphaBetaPlayer):
```

```
search_fn = 'minimax'
heuristic_fn = 'heuristic_liberties' # 'heuristic_liberties' /
'heuristic_area'
verbose_depth = True
def load(cls): pass
def save(cls): pass
play_sync( ( Agent(MiniMax, 'MiniMax'), Agent(AlphaBeta, 'AlphaBeta') ) )
pass
```

```
MiniMax
                     | depth: 1
AlphaBeta
                     | depth: 1 2
                     | depth: 1 2 3 4
MiniMax
AlphaBeta
                     | depth: 1 2 3 4 5
MiniMax
                     | depth: 1 2 3 4
AlphaBeta
                     | depth: 1 2 3 4 5
                     | depth: 1 2 3 4 5
MiniMax
                     | depth: 1 2 3 4 5
AlphaBeta
                     | depth: 1 2 3 4 5
MiniMax
                     | depth: 1 2 3 4 5 6
AlphaBeta
MiniMax
                     | depth: 1 2 3 4 5
AlphaBeta
                     | depth: 1 2 3 4 5 6
MiniMax
                     | depth: 1 2 3 4 5
AlphaBeta
                     | depth: 1 2 3 4 5 6
                     | depth: 1 2 3 4 5
MiniMax
                     | depth: 1 2 3 4 5
AlphaBeta
MiniMax
                     | depth: 1 2 3 4
                     | depth: 1 2 3 4 5
AlphaBeta
MiniMax
                     | depth: 1 2 3 4
                     | depth: 1 2 3 4 5
AlphaBeta
MiniMax
                     | depth: 1 2 3 4
                     | depth: 1 2 3 4 5 6
AlphaBeta
MiniMax
                     | depth: 1 2 3 4 5 6
AlphaBeta
                     | depth: 1 2 3 4 5 6 7
MiniMax
                     | depth: 1 2 3 4
AlphaBeta
                     | depth: 1 2 3 4 5
MiniMax
                     | depth: 1 2 3 4
                     | depth: 1 2 3 4 5
AlphaBeta
                     | depth: 1 2 3 4
MiniMax
                     | depth: 1 2 3 4 5
AlphaBeta
                     | depth: 1 2 3 4 5
MiniMax
AlphaBeta
                     | depth: 1 2 3 4 5
                     | depth: 1 2 3 4 5
MiniMax
AlphaBeta
                     | depth: 1 2 3 4 5
MiniMax
                     | depth: 1 2 3 4 5
AlphaBeta
                     | depth: 1 2 3 4 5 6 7
```

MiniMax | depth: 1 2 3 4 5 6 | depth: 1 2 3 4 5 6 7 AlphaBeta | depth: 1 2 3 4 5 6 MiniMax AlphaBeta | depth: 1 2 3 4 5 6 7 MiniMax | depth: 1 2 3 4 5 6 AlphaBeta | depth: 1 2 3 4 5 6 MiniMax | depth: 1 2 3 4 5 | depth: 1 2 3 4 5 6 7 AlphaBeta MiniMax | depth: 1 2 3 4 5 | depth: 1 2 3 4 5 6 AlphaBeta | depth: 1 2 3 4 5 6 MiniMax | depth: 1 2 3 4 5 6 7 AlphaBeta | depth: 1 2 3 4 5 6 7 MiniMax depth: 1 2 3 4 5 6 7 8 9 AlphaBeta | depth: 1 2 3 4 5 6 7 MiniMax AlphaBeta | depth: 1 2 3 4 5 6 7 8 9 MiniMax | depth: 1 2 3 4 5 6 7 8 AlphaBeta | depth: 1 2 3 4 5 6 7 8 9 MiniMax | depth: 1 2 3 4 5 6 7 8 9 10 AlphaBeta | depth: 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 inf MiniMax | depth: 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 -inf AlphaBeta | depth: 1 inf MiniMax | depth: 1 -inf

6 Monty Carlo Tree Search

- agents/MCTS.py
- agents/UCT.py

I have implemented variations on Monty Carlo Tree Search.

6.1 MCTSMaximum

is designed to be reinforcement learning agents, to be trained by running repeatedly before the match. Initially they will make random moves. After each match, the .backpropergate() function is called, which will build a record of the win/loss ratio for each seen board position, and also compute the BestChild score $(w/n + c\sqrt{(\ln N/n)})$ for each seen node. The agent is runtime fast in the sense that it only needs to read from it's own cache to compute the max score of available actions, which assumes the agent has seen this exact board position before during training,

6.2 MCTSRandom

Is a variation inspired by the Ant Colony Optimization Algorithm. It removes the exploration term, and instead of selecting the action with the maximum score, it uses the difference in score as random weighting factor to stochastically select the next action.

6.3 MCTSRandomHeuristic + MCTSMaximumHeuristic

These are variants that add the liberties heuristic to the BestChild score

6.4 Results

• MCTSRandom wins against MCTSMaximum by 68%

```
[25]: ! rm -f ./data/MCTS*.pickle
      ! python3 ./run_backpropagation.py -a MCR -o MCM -r 100 -l 100
                         1s | 70% -> 70% | MCTS Random vs MCTS Maximum
                100
     wrote:
             ./data/MCTSMaximum.zip.pickle
                                                      | 0.1MB in 0.1s | entries:
     4354
     wrote:
             ./data/MCTSRandom.zip.pickle
                                                        0.1MB in 0.1s | entries:
     4354
     MCTSRandom can be shown to be trainable against a Greedy opponent
[26]: | python3 ./run_backpropagation.py -a MCR -o GREEDY -r 1000
     loaded: ./data/MCTSRandom.zip.pickle
                                                        0.1MB in 0.0s | entries:
     4354
      match_id:
                 100
                               39% ->
                                      37% | MCTS Random vs Greedy
      match_id: 200 |
                               44% -> 46% | MCTS Random vs Greedy
      match_id: 300 |
                         4s |
                               47% -> 50% | MCTS Random vs Greedy
                         5s | 45% -> 46% | MCTS Random vs Greedy
      match_id: 400 |
      match_id: 500 |
                         7s |
                              45% -> 45% | MCTS Random vs Greedy
                         8s | 46% -> 47% | MCTS Random vs Greedy
      match_id: 600 |
      match id: 700 |
                         9s |
                              48% -> 49% | MCTS Random vs Greedy
      match id: 800 |
                        11s | 48% -> 49% | MCTS Random vs Greedy
      match id: 900 |
                        12s | 48% -> 50% | MCTS Random vs Greedy
      match id: 1000 |
                        13s |
                               48% -> 49% | MCTS Random vs Greedy
                                                      | 0.6MB in 0.8s | entries:
     wrote: ./data/MCTSRandom.zip.pickle
     29185
     MCTSRandom has a curious dip in performance against Minimax. - After training against Greedy,
```

MCTSRandom has a curious dip in performance against Minimax. - After training against Greedy, its performance is 17% - Repeatedly losing seems to amplify exploration, decreasing to a minimum 11% winrate - It eventually finds better countermoves, training itself back upto 18% winrate

```
[27]: | python3 ./run_backpropagation.py -a MCR -o MINIMAX -r 1000
     loaded: ./data/MCTSRandom.zip.pickle
                                                        0.6MB in 0.1s | entries:
     29185
                               17% ->
                                      17% | MCTS Random vs Minimax
      match_id:
                 100 |
                        32s |
      match id:
                        63s |
                               13% ->
                                      11% | MCTS Random vs Minimax
                 200
      match id:
                 300 l
                        92s |
                               13% -> 13% | MCTS Random vs Minimax
                               14% -> 14% | MCTS Random vs Minimax
      match id: 400 | 122s |
      match_id: 500 | 153s | 14% -> 14% | MCTS Random vs Minimax
```

```
match_id: 600 | 186s | 16% -> 17% | MCTS Random vs Minimax
match_id: 700 | 216s | 16% -> 18% | MCTS Random vs Minimax
match_id: 800 | 246s | 17% -> 19% | MCTS Random vs Minimax
match_id: 900 | 275s | 17% -> 18% | MCTS Random vs Minimax
match_id: 1000 | 303s | 17% -> 18% | MCTS Random vs Minimax
wrote: ./data/MCTSRandom.zip.pickle | 1.3MB in 1.8s | entries:
64980
```

It has a hard time against AlphaBeta, with only a 4-6% winrate

```
[28]: | python3 ./run_backpropagation.py -a MCR -o ALPHABETA -r 250 -1 50
```

```
1.3MB in 0.3s | entries:
loaded: ./data/MCTSRandom.zip.pickle
64980
match id:
            50 | 192s |
                          4% ->
                                 7% | MCTS Random vs AlphaBeta
                          6% -> 7% | MCTS Random vs AlphaBeta
match id: 100 | 377s |
match id: 150 | 570s |
                          6% ->
                                 6% | MCTS Random vs AlphaBeta
match id: 200 | 762s |
                         6% ->
                                 5% | MCTS Random vs AlphaBeta
match_id: 250 | 957s |
                          6% ->
                                 6% | MCTS Random vs AlphaBeta
wrote: ./data/AlphaBetaPlayer.zip.pickle
                                               | 4.7MB in 5.3s | entries:
308358
       ./data/MCTSRandom.zip.pickle
                                               | 1.5MB in 2.1s | entries:
wrote:
75992
```

What if we pretrain all the Monty Carlo agents against each other?

NOTE: Cross-training in a league is the method used by the Starcraft AlphaStar agent.

```
[29]: for agent in TEST_AGENTS.keys():
          if agent.startswith('MC'):
              TEST_AGENTS[agent].agent_class.verbose = False
              TEST_AGENTS[agent].agent_class.load()
      for agent in TEST_AGENTS.keys():
          for opponent in TEST_AGENTS.keys():
              if agent.startswith('MC') and opponent.startswith('MC'):
                  time.sleep(0.1)
                  run backpropagation({
                      "agent":
                                    agent,
                      "opponent":
                                    opponent,
                      "rounds":
                                    1000,
                      "logging":
                                    1000,
                      "progress":
                                    False,
                      "time_limit": 0  # reduce freak TimeoutErrors
                  })
      for agent in TEST_AGENTS.keys():
          if agent.startswith('MC'):
              TEST_AGENTS[agent].agent_class.save()
```

```
51% ->
                                  50% | MCTS Maximum vs MCTS Maximum 2
match_id: 1000 |
                    9s |
match_id: 1000 |
                          26% ->
                    9s |
                                  26% | MCTS Maximum vs MCTS Random
match_id: 1000 |
                   10s |
                          53% ->
                                  51% | MCTS Maximum vs MCTS Maximum Heuristic
match id: 1000 |
                   20s |
                          31% ->
                                  32% | MCTS Maximum vs MCTS Random Heuristic
match id: 1000 |
                          70% ->
                                  72% | MCTS Random vs MCTS Maximum
                   11s |
match id: 1000 |
                   10s |
                          49% ->
                                  49% | MCTS Random vs MCTS Random 2
match id: 1000 |
                   12s |
                          74% ->
                                  74% | MCTS Random vs MCTS Maximum Heuristic
match_id: 1000 |
                   14s |
                          53% ->
                                  55% | MCTS Random vs MCTS Random Heuristic
match id: 1000 |
                          53% -> 51% | MCTS Maximum Heuristic vs MCTS Maximum
                   10s |
match id: 1000 |
                          24% -> 24% | MCTS Maximum Heuristic vs MCTS Random
                   12s |
match_id: 1000 |
                   12s |
                          45% -> 46% | MCTS Maximum Heuristic vs MCTS Maximum
Heuristic 2
                          32% ->
 match_id: 1000 | 17s |
                                  32% | MCTS Maximum Heuristic vs MCTS Random
Heuristic
match_id: 1000 |
                   22s |
                          69% ->
                                  70% | MCTS Random Heuristic vs MCTS Maximum
match_id: 1000 |
                   15s |
                          46% ->
                                  47% | MCTS Random Heuristic vs MCTS Random
match_id: 1000 |
                   17s |
                          67% ->
                                  65% | MCTS Random Heuristic vs MCTS Maximum
Heuristic
match id: 1000 |
                          48% -> 48% | MCTS Random Heuristic vs MCTS Random
                   22s |
Heuristic 2
```

Then attempt a rematch against AlphaBeta, which results in a small but significant winrate improvement.

```
[30]: | python3 ./run_backpropagation.py -a MCR -o ALPHABETA -r 250 -1 50
                                                      | 4.7MB in 1.4s | entries:
     loaded: ./data/AlphaBetaPlayer.zip.pickle
     308358
                                                        7.2MB in 2.1s | entries:
     loaded: ./data/MCTSRandom.zip.pickle
     369486
      match id:
                                4% ->
                                        5% | MCTS Random vs AlphaBeta
                  50 | 194s |
                                        4% | MCTS Random vs AlphaBeta
      match id: 100 | 378s |
                                4% ->
      match id:
                150 | 565s |
                                5% ->
                                        5% | MCTS Random vs AlphaBeta
      match_id: 200 | 752s |
                                5% ->
                                        5% | MCTS Random vs AlphaBeta
      match id:
                 250 | 947s |
                                5% ->
                                        5% | MCTS Random vs AlphaBeta
                                                      | 9.5MB in 12.0s | entries:
     wrote:
             ./data/AlphaBetaPlayer.zip.pickle
     629184
             ./data/MCTSRandom.zip.pickle
                                                      | 7.4MB in 10.1s | entries:
     wrote:
     380447
```

6.5 UCTPlayer

UCTPlayer will use its 150ms of time to simulate MCTSMaximum from the current board position before returning an answer based on the current scores of current available actions.

```
[31]: | python3 ./run_backpropagation.py -a UCT -o GREEDY -r 70 --progress
```

```
loaded: ./data/MCTSMaximum.zip.pickle | 5.4MB in 1.5s | entries:
    280851
    loaded: ./data/MCTSMaximum.zip.pickle | 5.4MB in 1.5s | entries:
    280851
    70 | 266s | 24% -> 24% | UCT vs Greedy
                                   | 5.7MB in 9.0s | entries:
    wrote: ./data/MCTSMaximum.zip.pickle
    295859
[32]: | python3 ./run_backpropagation.py -a UCT -o MINIMAX -r 70 --progress
    loaded: ./data/MCTSMaximum.zip.pickle | 5.7MB in 2.0s | entries:
    295859
    loaded: ./data/MCTSMaximum.zip.pickle | 5.7MB in 2.0s | entries:
    -----+- match id:
    70 | 282s | 3% -> 4% | UCT vs Minimax
    wrote: ./data/MCTSMaximum.zip.pickle
                                      | 6.0MB in 8.3s | entries:
    307631
[33]: | python3 ./run_backpropagation.py -a UCT -o ALPHABETA -r 70 --progress
    loaded: ./data/AlphaBetaPlayer.zip.pickle | 9.5MB in 2.8s | entries:
    629184
                                   | 6.0MB in 1.8s | entries:
    loaded: ./data/MCTSMaximum.zip.pickle
    307631
    loaded: ./data/MCTSMaximum.zip.pickle | 6.0MB in 1.7s | entries:
    307631
    -----+ match id:
    70 | 529s | 4% -> 5% | UCT vs AlphaBeta
    wrote: ./data/MCTSMaximum.zip.pickle | 6.2MB in 8.8s | entries:
    319067
         ./data/AlphaBetaPlayer.zip.pickle | 10.8MB in 16.7s | entries:
    wrote:
    719376
```

7 League Tables

Lets run every agent against every other agent and compare results - AlphaBetaArea has a 100% winrate against everything except AlphaBeta and MCTS Random - AlphaBetaArea vs AlphaBeta scores 60/40 both ways, which may depend on who has the bigger cache - MCTS Random scored a maximum 90% vs Alphabeta and 24% vs AlphaBeta Area in reverse matchups, - MCTS Random got beaten 100% Alphabeta and AlphaBeta later when in was in first position, this may be cache related

```
[3]: for agent in TEST_AGENTS.keys():
    for opponent in TEST_AGENTS.keys():
        try: TEST_AGENTS[agent].agent_class.load()
```

```
except: pass
        time.sleep(0.1)
        TEST_AGENTS[agent
                           ].agent_class.verbose = False
        TEST_AGENTS[opponent].agent_class.verbose = False
        is_slow = any(
            name in TEST_AGENTS[agent].name + TEST_AGENTS[opponent].name
            for name in ['Alpha', 'Area', 'UCT', 'Minimax', 'Custom']
        )
        run_backpropagation({
            "agent":
                         agent,
            "opponent":
                         opponent,
            "time limit": 150,
            "rounds":
                         10 if is_slow else 100,
            "progress":
                         False,
            "exceptions": False,
       })
   print()
for agent in TEST_AGENTS.keys():
   try: TEST_AGENTS[agent].agent_class.save()
    except: pass
                         52% ->
                                 54% | Random vs Random 2
match_id:
           100
                   1s |
                         17% ->
match_id:
           100
                   1s |
                                 17% | Random vs Greedy
                   1s | 61% -> 62% | Random vs Distance
match id:
           100 l
match_id:
          100
                   2s |
                         22% ->
                                21% | Random vs Greedy Distance
                   3s l
                         0% ->
match id:
            10 l
                                 0% | Random vs Minimax
                  41s | 30% -> 23% | Random vs AlphaBeta
match id:
            10 |
match id:
            10 l
                  42s |
                         0% ->
                                  0% | Random vs AlphaBeta Area
match_id: 100 |
                   2s | 70% -> 69% | Random vs MCTS Maximum
                   4s l
match id:
          100 l
                        50% -> 48% | Random vs MCTS Random
match_id:
          100 |
                   2s l
                        70% -> 69% | Random vs MCTS Maximum Heuristic
                        58% -> 62% | Random vs MCTS Random Heuristic
match_id:
          100
                   1s |
match_id:
            10 |
                  35s |
                          0% ->
                                  0% | Random vs UCT
                  35s |
                          0% ->
                                  0% | Random vs Custom TestAgent
match_id:
            10
match_id:
           100
                   1s |
                         75% ->
                                 75% | Greedy vs Random
                   1s l
                         50% ->
                                 50% | Greedy vs Greedy 2
match_id:
           100 l
match_id:
          100 |
                   2s | 50% -> 50% | Greedy vs Distance
match_id:
          100
                   2s |
                        50% -> 50% | Greedy vs Greedy Distance
            10 l
                   3s |
                         0% ->
                                  0% | Greedy vs Minimax
match_id:
match id:
            10 l
                  39s | 10% ->
                                  1% | Greedy vs AlphaBeta
match id:
                         0% ->
                                 0% | Greedy vs AlphaBeta Area
            10 |
                  31s |
match id: 100 |
                   1s | 87% -> 89% | Greedy vs MCTS Maximum
match id:
           100 l
                   1s | 67% ->
                                 64% | Greedy vs MCTS Random
                   1s | 91% -> 90% | Greedy vs MCTS Maximum Heuristic
match id:
          100
```

```
83% -> 82% | Greedy vs MCTS Random Heuristic
 match_id:
            100
                    2s |
                   36s | 100% -> 100% | Greedy vs UCT
match_id:
             10
 match_id:
             10 l
                   37s |
                           0% ->
                                   0% | Greedy vs Custom TestAgent
                          40% ->
 match id:
            100 l
                    1s l
                                   39% | Distance vs Random
 match id:
            100 l
                    1s l
                          50% ->
                                  50% | Distance vs Greedy
 match id:
            100
                    2s |
                          50% ->
                                   50% | Distance vs Distance 2
 match id:
            100 l
                    2s l
                           0% ->
                                   0% | Distance vs Greedy Distance
                           0% ->
match id:
             10 |
                    2s |
                                   0% | Distance vs Minimax
match_id:
             10 l
                   51s |
                           0% ->
                                   0% | Distance vs AlphaBeta
                           0% ->
 match_id:
             10 |
                   60s |
                                   0% | Distance vs AlphaBeta Area
                          50% ->
                                  49% | Distance vs MCTS Maximum
 match_id:
            100
                    2s |
 match_id:
                    2s |
                          31% ->
                                  29% | Distance vs MCTS Random
            100
                          22% ->
 match_id:
            100
                    2s |
                                  17% | Distance vs MCTS Maximum Heuristic
 match_id:
            100
                    2s |
                          46% ->
                                  48% | Distance vs MCTS Random Heuristic
match_id:
                   40s l
                          50% ->
                                  41% | Distance vs UCT
             10
 match_id:
             10 |
                   48s |
                           0% ->
                                   0% | Distance vs Custom TestAgent
                          87% ->
                                  90% | Greedy Distance vs Random
match_id:
            100 |
                    2s |
 match id:
            100 l
                    2s l
                          50% ->
                                  50% | Greedy Distance vs Greedy
                    2s | 100% -> 100% | Greedy Distance vs Distance
 match id:
            100 l
                    2s |
match id:
            100
                          50% ->
                                  50% | Greedy Distance vs Greedy Distance 2
match id:
             10 l
                    3s l
                          50% ->
                                  45% | Greedy Distance vs Minimax
                   38s |
                          20% ->
                                  10% | Greedy Distance vs AlphaBeta
match_id:
             10 l
match_id:
             10 |
                   35s |
                           0% ->
                                   0% | Greedy Distance vs AlphaBeta Area
                          65% ->
match_id:
            100 |
                    2s |
                                  68% | Greedy Distance vs MCTS Maximum
                    2s |
                          81% ->
                                  80% | Greedy Distance vs MCTS Random
 match_id:
            100
 match_id:
            100
                    2s |
                          82% ->
                                  76% | Greedy Distance vs MCTS Maximum
Heuristic
match_id:
            100 |
                          82% ->
                                  79% | Greedy Distance vs MCTS Random Heuristic
                    2s |
match id:
             10 l
                   33s l
                          90% ->
                                  99% | Greedy Distance vs UCT
match_id:
             10
                   44s |
                          10% ->
                                  17% | Greedy Distance vs Custom TestAgent
             10 |
                    3s | 100% -> 100% | Minimax vs Random
match_id:
                    3s | 100% -> 100% | Minimax vs Greedy
match id:
             10 |
match id:
             10 l
                    2s | 100% -> 100% | Minimax vs Distance
                    3s | 100% -> 100% | Minimax vs Greedy Distance
match id:
             10 |
 match id:
             10 l
                          50% ->
                                  45% | Minimax vs Minimax 2
match_id:
                          20% ->
                                  24% | Minimax vs AlphaBeta
             10 |
                   44s |
match_id:
             10 l
                   42s |
                           0% ->
                                   0% | Minimax vs AlphaBeta Area
match_id:
                    2s | 100% -> 100% | Minimax vs MCTS Maximum
             10 |
             10 |
                   10s | 100% -> 100% | Minimax vs MCTS Random
 match_id:
match_id:
             10 |
                    2s | 100% -> 100% | Minimax vs MCTS Maximum Heuristic
                    3s | 100% -> 100% | Minimax vs MCTS Random Heuristic
match id:
             10 l
 match_id:
             10 |
                   37s | 100% -> 100% | Minimax vs UCT
 match_id:
                   47s |
                           0% ->
                                   0% | Minimax vs Custom TestAgent
                   35s | 100% -> 100% | AlphaBeta vs Random
match_id:
             10 |
```

```
37s | 90% -> 85% | AlphaBeta vs Greedy
match_id:
            10
match_id:
                  46s | 100% -> 100% | AlphaBeta vs Distance
match_id:
            10 |
                  38s | 100% -> 100% | AlphaBeta vs Greedy Distance
                        80% -> 92% | AlphaBeta vs Minimax
match_id:
            10 |
                  43s |
match id:
            10 l
                  96s l
                         40% ->
                                 38% | AlphaBeta vs AlphaBeta 2
match id:
                         20% -> 24% | AlphaBeta vs AlphaBeta Area
            10 |
match id:
                  25s | 100% -> 100% | AlphaBeta vs MCTS Maximum
match id:
            10 l
                  38s |
                        90% -> 95% | AlphaBeta vs MCTS Random
match id:
            10 |
                  31s | 100% -> 100% | AlphaBeta vs MCTS Maximum Heuristic
match_id:
            10 l
                  36s | 100% -> 100% | AlphaBeta vs MCTS Random Heuristic
                  91s | 100% -> 100% | AlphaBeta vs UCT
match_id:
            10 |
                  86s | 40% -> 38% | AlphaBeta vs Custom TestAgent
match_id:
            10 |
match_id:
            10
                  34s | 100% -> 100% | AlphaBeta Area vs Random
match_id:
                  31s | 100% -> 100% | AlphaBeta Area vs Greedy
                  43s | 100% -> 100% | AlphaBeta Area vs Distance
match_id:
match_id:
            10 |
                  34s | 100% -> 100% | AlphaBeta Area vs Greedy Distance
match_id:
            10 | 127s | 100% -> 100% | AlphaBeta Area vs Minimax
match_id:
                  89s | 40% -> 64% | AlphaBeta Area vs AlphaBeta
            10 l
match id:
                  82s |
                        60% -> 52% | AlphaBeta Area vs AlphaBeta Area 2
            10 l
                  30s | 100% -> 100% | AlphaBeta Area vs MCTS Maximum
match id:
match id:
            10 | 313s | 80% -> 76% | AlphaBeta Area vs MCTS Random
match id:
                  30s | 100% -> 100% | AlphaBeta Area vs MCTS Maximum Heuristic
                  37s | 100% -> 100% | AlphaBeta Area vs MCTS Random Heuristic
match_id:
            10 |
match_id:
            10 |
                  75s | 100% -> 100% | AlphaBeta Area vs UCT
                        50% -> 45% | AlphaBeta Area vs Custom TestAgent
match_id:
            10 |
                  90s |
                         34% ->
match_id:
           100
                   2s |
                                 34% | MCTS Maximum vs Random
                   2s |
                         13% ->
match id:
           100
                                 15% | MCTS Maximum vs Greedy
match_id:
           100
                   2s |
                         58% ->
                                 57% | MCTS Maximum vs Distance
                   2s |
                         19% -> 28% | MCTS Maximum vs Greedy Distance
match_id:
          100 l
match_id:
            10 |
                  10s |
                          0% ->
                                  0% | MCTS Maximum vs Minimax
match_id:
            10
                  24s |
                          0% ->
                                  0% | MCTS Maximum vs AlphaBeta
                          0% ->
                                  0% | MCTS Maximum vs AlphaBeta Area
match_id:
            10 |
                  36s |
match id:
          100
                   2s |
                         52% ->
                                 51% | MCTS Maximum vs MCTS Maximum 2
match id:
           100 l
                   3s l
                         29% ->
                                 36% | MCTS Maximum vs MCTS Random
match id:
           100
                   2s l
                         46% ->
                                 41% | MCTS Maximum vs MCTS Maximum Heuristic
match id:
           100 l
                   2s l
                         36% ->
                                 33% | MCTS Maximum vs MCTS Random Heuristic
                         40% ->
                                 44% | MCTS Maximum vs UCT
match_id:
            10 l
                  34s |
                          0% ->
match_id:
            10 l
                  33s |
                                  0% | MCTS Maximum vs Custom TestAgent
                         46% ->
           100
                   1s |
                                 48% | MCTS Random vs Random
\mathtt{match\_id}:
match_id:
           100
                   1s |
                         32% ->
                                 34% | MCTS Random vs Greedy
match_id:
           100
                   2s |
                         64% ->
                                 67% | MCTS Random vs Distance
match_id:
           100
                   2s |
                         18% -> 16% | MCTS Random vs Greedy Distance
match_id:
            10
                   4s l
                          0% ->
                                  0% | MCTS Random vs Minimax
match_id:
            10 | 425s |
                          0% ->
                                  0% | MCTS Random vs AlphaBeta
match_id:
                          0% ->
                                  0% | MCTS Random vs AlphaBeta Area
            10 |
                  56s |
```

```
77% ->
                                   81% | MCTS Random vs MCTS Maximum
 match_id:
            100
                    2s l
                           44% ->
match_id:
            100
                    2s
                                   44% | MCTS Random vs MCTS Random 2
 match_id:
            100
                    2s l
                           74% ->
                                   75% | MCTS Random vs MCTS Maximum Heuristic
 match_id:
            100 |
                    2s |
                           48% ->
                                   49% | MCTS Random vs MCTS Random Heuristic
 match id:
                                   39% | MCTS Random vs UCT
             10 l
                    32s |
                           50% ->
 match id:
                    47s |
                            0% ->
                                    0% | MCTS Random vs Custom TestAgent
             10 l
match_id:
            100 l
                    1s l
                           29% ->
                                   27% | MCTS Maximum Heuristic vs Random
match id:
            100 l
                           14% ->
                                   13% | MCTS Maximum Heuristic vs Greedy
                           64% ->
match_id:
            100 l
                    2s l
                                   69% | MCTS Maximum Heuristic vs Distance
 match_id:
                    2s l
                           35% ->
                                   34% | MCTS Maximum Heuristic vs Greedy
            100 |
Distance
                            0% ->
 match_id:
             10 |
                    6s |
                                    0% | MCTS Maximum Heuristic vs Minimax
                    34s |
                            0% ->
                                    0% | MCTS Maximum Heuristic vs AlphaBeta
match_id:
             10
                    32s |
                            0% ->
match_id:
             10
                                    0% | MCTS Maximum Heuristic vs AlphaBeta Area
match_id:
            100 l
                    2s l
                           48% ->
                                   46% | MCTS Maximum Heuristic vs MCTS Maximum
match_id:
            100 |
                    2s |
                           35% ->
                                   33% | MCTS Maximum Heuristic vs MCTS Random
match_id:
            100 |
                    2s |
                           46% ->
                                   50% | MCTS Maximum Heuristic vs MCTS Maximum
Heuristic 2
 match id:
            100 l
                    2s l
                           35% ->
                                   36% | MCTS Maximum Heuristic vs MCTS Random
Heuristic
match id:
                                   52% | MCTS Maximum Heuristic vs UCT
             10 |
                    32s |
                           40% ->
match id:
                    31s |
                            0% ->
                                    0% | MCTS Maximum Heuristic vs Custom
TestAgent
```

8 Opening Book

8.1 Monty Carlo Method

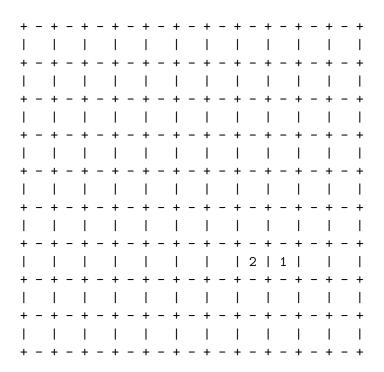
We can derive an opening book by looking at the cached scores in MCTSR andom - score priorities exploration of 100% winrate nodes that have only been explored once - sorting by score actually suggested the best countermove was the corner square - sort by wins * wins/count to find the best move

- Monty Carlo suggests the best opening strategy is:
 - defend the corner on the 3,3 point
 - attack the knights blind spot on the 4,3 point
 - * knight requires three turns to move one space sideways
 - the first player is attempting to move into position to directly attack their opponent
 - the second player is attempting to remain in the blind spot adjcent to the first player
 - the pieces only diverge after move 8

```
[4]: from agents.MCTS import MCTSRandom from isolation.isolation import Isolation, DebugState MCTSRandom.load()
```

```
def best_response(parent_state=Isolation(), verbose=True):
              = parent_state.actions()
   actions
    child_states = [ parent_state.result(action) for action in actions ]
    child_states = [ child for child in child_states if child in MCTSRandom.
 data 1 -
   records
              = [ MCTSRandom.data[child] for child in child states ]
   best_state, best_record = max(zip(child_states, records),
                             key=lambda key_value: key_value[1].wins *_
 →key_value[1].wins/key_value[1].count)
              = DebugState.ind2xy(best_state.locs[parent_state.player()])
   best move
    if verbose:
       print('Best move '+str(best state.ply count)+' is:')
       print(best_move, best_record)
       print( DebugState.from_state(best_state).bitboard_string )
       print( DebugState.from_state(best_state) )
   return best_move, best_record, best_state
best_state = Isolation()
for i in range(8):
   try:
       best_move, best_record, best_state = best_response(best_state)
   except: pass
loaded: ./data/MCTSRandom.zip.pickle
                                 | 7.4MB in 2.6s | entries:
380447
Best move 1 is:
(2, 2) MCTSRecord(wins=977, count=1487, score=0.6567967698519516)
111111011001111111111110011111111111
```

Best move 2 is:

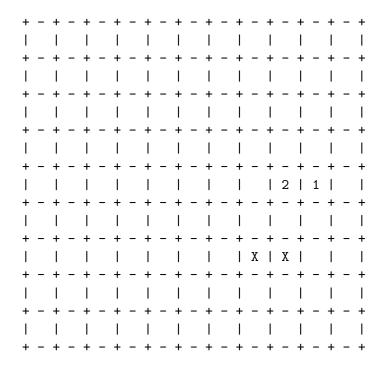


Best move 3 is:

+	-	+	-	+	-	+	-	+	-	+	-	+	-	+	-	+	-	+	-	+	-	+
				-						-								-				
+	-	+	-	+	-	+	-	+	-	+	-	+	-	+	-	+	-	+	-	+	-	+
		-		-				-		-		-		-		-		-		-		
+	-	+	-	+	-	+	-	+	-	+	-	+	-	+	-	+	-	+	-	+	-	+
				-						-												
+	-	+	-	+	-	+	-	+	-	+	_	+	_	+	-	+	-	+	-	+	-	+
		-		-		-		-		1		-		-		-				-		1
+	-	+	_	+	-	+	-	+	-	+	-	+	-	+	-	+	-	+	-	+	-	+
				-						1						-		-	1	-		
+	-	+	_	+	-	+	-	+	-	+	-	+	-	+	-	+	-	+	-	+	-	+
				-						1						-		-		-		
+	-	+	_	+	-	+	-	+	-	+	-	+	-	+	-	+	-	+	-	+	-	+
		-		-		-		-		1		-		-	2	-	Х	1		-		
_	_	_	_	_	_	_	_	_	_	_	_	_	_	_	_	_	_	_	_	_	_	_

_		_	_	_	_	_	_	_	_	_	_	_	_	_	_	_	_	_	_	+	_	_
																		-		-		
+	-	+	_	+	-	+	-	+	-	+	-	+	-	+	-	+	-	+	-	+	-	+

Best move 4 is:

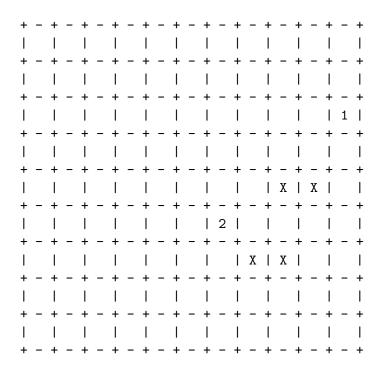


Best move 5 is:

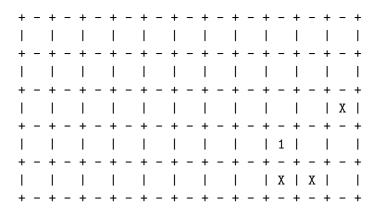
+	-	+	-	+	-	+	-	+	-	+	-	+	-	+	-	+	-	+	-	+	-	+
		-														-		-				
+	-	+	_	+	-	+	-	+	-	+	-	+	-	+	-	+	-	+	_	+	-	+
		-		-		-		-		-		-		-		-		-		-		
+	-	+	-	+	-	+	-	+	-	+	-	+	-	+	-	+	-	+	_	+	-	+
1		-		-		-		1		1		-		-		1		1		-	1	
+	-	+	-	+	-	+	-	+	-	+	-	+	_	+	-	+	_	+	_	+	-	+
		1		-		-		1		1		-		-		1		1		1		
+	-	+	_	+	-	+	-	+	-	+	-	+	-	+	-	+	-	+	_	+	-	+
		-		-		-		-		-		-		-		-	2	-	X	-		
+	-	+	-	+	-	+	-	+	-	+	-	+	_	+	-	+	_	+	_	+	-	+
		1		1		1		1		1		1		1		1		1		1		-
+	_	+	_	+	_	+	_	+	_	+	_	+	_	+	_	+	_	+	_	+	_	+

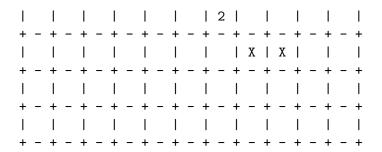
															X		X					
+	-	+	-	+	-	+	-	+	-	+	-	+	-	+	-	+	-	+	-	+	-	+
																-						-
+	_	+	_	+	-	+	-	+	-	+	-	+	-	+	-	+	-	+	-	+	-	+
				-												-		-				-
+	_	+	_	+	_	+	_	+	_	+	_	+	_	+	_	+	_	+	_	+	_	+

Best move 6 is:

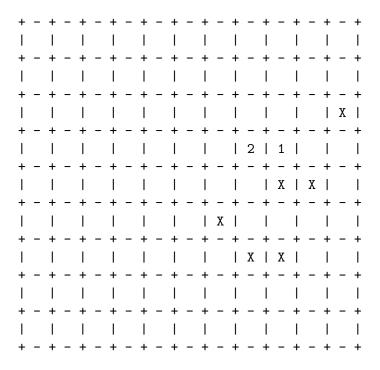


Best move 7 is:





Best move 8 is:



8.2 AlphaBetaArea Method

Alteratively we could see how AlphaBetaArea plays when given different depths to search during the opening

Opening Move: - Monty Carlo: likes the 3,3 corner position - Depth 1: main diagonal, one off from center - Depth 2: short centerline, one off from center - Depth 3: center,3 position - Depth 4: corner 4,4 position - Depth 5: long-edge but one off from center

Subsequent Moves: - Usually ends up in either an attacking or adjacent square - Except at depth 4, when p1 approaches diagonally, p2 jumps away

```
[5]: AlphaBetaAreaPlayer.load()
     class AlphaBetaOpening(AlphaBetaAreaPlayer):
         data
                           = AlphaBetaAreaPlayer.data # import cache
         verbose_depth
     for depth in range(1,6):
         AlphaBetaOpening.search_max_depth = depth
         play_sync(
             agents=(
                 Agent(AlphaBetaOpening, 'AlphaBetaOpening'),
                 Agent(AlphaBetaOpening, 'AlphaBetaOpening'),
             ),
             verbose=True,
             time_limit=0,
             max_moves=4,
         )
    AlphaBetaOpening
                          | depth: 1
    AlphaBetaOpening
                        | depth: 1
    match: 0 \mid move: 2 \mid 0.06s \mid AlphaBetaOpening(1) \Rightarrow (6, 4)
```

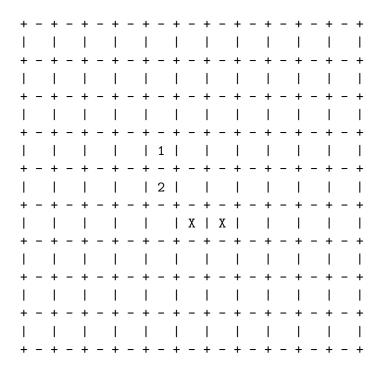
```
AlphaBetaOpening | depth: 1
AlphaBetaOpening
             | depth: 1
match: 0 | move: 4 | 0.01s | AlphaBetaOpening(1) => (11, 0)
```

```
AlphaBetaOpening | depth: 1 2
AlphaBetaOpening | depth: 1 2
```

match: $0 \mid move: 2 \mid 0.73s \mid AlphaBetaOpening(1) \Rightarrow (4, 3)$

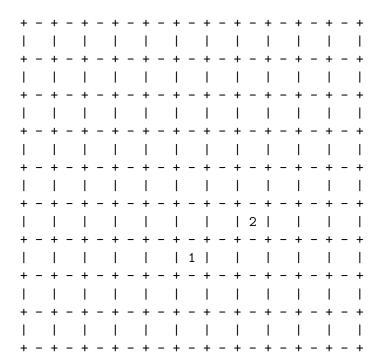
```
AlphaBetaOpening | depth: 1 2
AlphaBetaOpening | depth: 1 2
```

match: 0 | move: 4 | 0.04s | AlphaBetaOpening(1) => (2, 1)



AlphaBetaOpening | depth: 1 2 3 AlphaBetaOpening | depth: 1 2 3

match: $0 \mid move: 2 \mid 1.65s \mid AlphaBetaOpening(1) \Rightarrow (3, 3)$



AlphaBetaOpening | depth: 1 2 3

| depth: 1 2 3 AlphaBetaOpening match: $0 \mid move: 4 \mid 0.11s \mid AlphaBetaOpening(1) \Rightarrow (1, -2)$ | X | | 2 | 1 | AlphaBetaOpening | depth: 1 2 3 4 AlphaBetaOpening | depth: 1 2 3 4 match: 0 | move: 2 | 9.82s | AlphaBetaOpening(1) => (5, 2) | 1 | | 2 |

AlphaBetaOpening | depth: 1 2 3 4 AlphaBetaOpening | depth: 1 2 3 4 match: 0 | move: 4 | 0.28s | AlphaBetaOpening(1) => (2, 1) | X | AlphaBetaOpening | depth: 1 2 3 4 5 | depth: 1 2 3 4 5 AlphaBetaOpening match: $0 \mid move: 2 \mid 32.56s \mid AlphaBetaOpening(1) \Rightarrow (5, 2)$

| 1 |

- 1

9 Full Game (Final Section)

Here is a full game between AlphaBeta and AlphaBetaArea, with 1 second per move - Curiously, AlphaBeta at depth 2 picks the Monty Carlo 3,3 opening position - AlphaBetaArea(2) figures out forced checkmate in 26 on move 54 delivers it on move 65 (11 moves later) - The checkmate strategy for player 2 involves forcing 1 into a trapped corner

```
[18]: AlphaBetaPlayer.verbose_depth = True
AlphaBetaAreaPlayer.verbose_depth = True
play_sync(
    agents=(
        Agent(AlphaBetaPlayer, 'AlphaBeta'),
        Agent(AlphaBetaAreaPlayer, 'AlphaBetaArea'),
    ),
    verbose=True,
    verbose_depth=True,
    time_limit=1*1000,
)
```

```
AlphaBetaPlayer.verbose_depth = False
AlphaBetaAreaPlayer.verbose_depth = False
```

AlphaBetaPlayer | depth: 1 2 AlphaBetaAreaPlayer | depth: 1 2 3 match: 0 | move: 2 | 1.00s | AlphaBetaAreaPlayer(1) => (6, 2) AlphaBetaPlayer | depth: 1 2 3 4 5 6 7 AlphaBetaAreaPlayer | depth: 1 2 3 4 5 match: $0 \mid move: 4 \mid 1.00s \mid AlphaBetaAreaPlayer(1) \Rightarrow (2, 1)$ | X | | X |

```
AlphaBetaPlayer
             | depth: 1 2 3 4 5 6 7
AlphaBetaAreaPlayer | depth: 1 2 3 4 5 6
match: 0 | move: 6 | 1.00s | AlphaBetaAreaPlayer(1) => (12, 1)
     | | 2 |
  AlphaBetaPlayer
             | depth: 1 2 3 4 5 6 7 8
AlphaBetaAreaPlayer | depth: 1 2 3 4 5
match: 0 | move: 8 | 1.00s | AlphaBetaAreaPlayer(1) => (11, 0)
           | 2 |
```

```
- 1
                        | X |
AlphaBetaPlayer
                   | depth: 1 2 3 4 5 6 7
AlphaBetaAreaPlayer | depth: 1 2 3 4 5
match: 0 | move: 10 | 1.01s | AlphaBetaAreaPlayer(1) => (11, -2)
                  | X |
              | X |
                             | X |
AlphaBetaPlayer
                   | depth: 1 2 3 4 5 6 7
AlphaBetaAreaPlayer | depth: 1 2 3 4 5 6 7
match: 0 | move: 12 | 1.00s | AlphaBetaAreaPlayer(1) => (12, -3)
```

```
| | X | | | | | 2 |
  | X | |
             | depth: 1 2 3 4 5 6 7
AlphaBetaPlayer
AlphaBetaAreaPlayer | depth: 1 2 3 4 5 6
match: 0 | move: 14 | 1.00s | AlphaBetaAreaPlayer(1) => (2, 1)
          | X |
  | | X | | | |
                    | X | 1 | |
  | depth: 1 2 3 4 5 6 7
AlphaBetaPlayer
AlphaBetaAreaPlayer | depth: 1 2 3 4 5
match: 0 | move: 16 | 1.01s | AlphaBetaAreaPlayer(1) => (12, 1)
+ - + - + - + - + - + - + - +
```

```
| X |
           | depth: 1 2 3 4 5 6 7
AlphaBetaPlayer
AlphaBetaAreaPlayer | depth: 1 2 3 4 5
match: 0 | move: 18 | 1.00s | AlphaBetaAreaPlayer(1) => (11, -2)
| | X |
                | X |
           | depth: 1 2 3 4 5 6 7 8
AlphaBetaPlayer
AlphaBetaAreaPlayer | depth: 1 2 3 4 5 6
match: 0 | move: 20 | 1.00s | AlphaBetaAreaPlayer(1) => (2, -1)
```

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| X | X | X | X |
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        | X |
                      | X | X |
           | X |
                   | 1 | X |
                      | X |
 | X |
                      AlphaBetaPlayer
              | depth: 1 2 3 4 5 6 7
AlphaBetaAreaPlayer | depth: 1 2 3 4 5 6
match: 0 | move: 22 | 1.00s | AlphaBetaAreaPlayer(1) => (2, -1)
        | | X | | |
                   | X | X | X | X |
               | X | X | 1 | X |
     | X | | 2 |
           | X |
                   | X | X |
| | | | | | | | | | | | | | |
+-+-+-+-+
              | depth: 1 2 3 4 5 6 7
AlphaBetaPlayer
AlphaBetaAreaPlayer | depth: 1 2 3 4 5
match: 0 | move: 24 | 1.01s | AlphaBetaAreaPlayer(1) => (2, -1)
```

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-+-+-+-+
     | | X | | | X | X | X | X |
                | X | X | X | X |
             | X | 1 |
                      | X | X |
                  | X | X |
        | 2 | X |
AlphaBetaPlayer
              | depth: 1 2 3 4 5 6 7
AlphaBetaAreaPlayer | depth: 1 2 3 4 5 6
match: 0 | move: 26 | 1.00s | AlphaBetaAreaPlayer(1) => (12, 1)
           | | X | | X | |
             | 2 |
               | X | X | X | X |
                      | X | X |
  | | X |
           | | X | X |
     | X | X |
             | 1 |
           AlphaBetaPlayer
              | depth: 1 2 3 4 5 6 7 8
AlphaBetaAreaPlayer | depth: 1 2 3 4 5 6
match: 0 | move: 28 | 1.00s | AlphaBetaAreaPlayer(1) => (2, 1)
```

```
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 | | 2 | X |
     - 1
        | X | X | X | X | | | | | | |
   | | X |
       | X | X | X | X |
| | X | | | X | X | | X | X | |
| X | X |
| | X | | X |
+ - + - + - + - + - + - + - + - + - +
```

AlphaBetaPlayer | depth: 1 2 3 4 5 6 7 8 9

AlphaBetaAreaPlayer | depth: 1 2 3 4 5 6

match: 0 | move: 30 | 1.00s | AlphaBetaAreaPlayer(1) => (11, 0)

```
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   | | X | X |
    | | X | | | X | X | | X | X | |
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| | 1 | | | | | | | | | | | | | | |
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```

```
AlphaBetaPlayer | depth: 1 2 3 4 5 6 7 8 9
```

AlphaBetaAreaPlayer | depth: 1 2 3 4 5

match: 0 | move: 32 | 1.00s | AlphaBetaAreaPlayer(1) => (1, -2)

```
| X | X |
         | X |
-+-+-+-+
   | X | X |
        1
         | X | X | X | X |
  | |2|X| |X|X|X| | |
+ - + - + - + - + - + - + - + - + - +
| 1 | X | | X | X |
            | X | X | |
+ - + - + - + - + - + - + - + - + - +
| X | | | X | | | X | |
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```

AlphaBetaPlayer | depth: 1 2 3 4 5 6 7 8

AlphaBetaAreaPlayer | depth: 1 2 3 4 5

match: 0 | move: 34 | 1.11s | AlphaBetaAreaPlayer(1) => (1, 2)

```
I \quad I \quad I \quad I
         1 1
   |2||X|X|
           | X |
| X | X |
        + - + - + - + - + - + - + - + - + - +
 | | | X | X | | X | X | X | X | |
+ - + - + - + - + - + - + - + - + - +
+-+-+-+-+
+ - + - + - + - + - + - + - + - + - +
| | X | | | X | | | X | | |
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+ - + - + - + - + - + - + - + - + - +
```

AlphaBetaPlayer | depth: 1 2 3 4 5 6 7 8

AlphaBetaAreaPlayer | depth: 1 2 3 4 5

```
match: 0 | move: 36 | 1.02s | AlphaBetaAreaPlayer(1) => (1, -2)
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 1
         | X |
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      | X | X |
          1 X I
   |X|
+ - + - + - + - + - + - + - + - + - +
 | | X | X | | | | X | X | X | X |
| 2 | | X | X | | X | X | X | X | |
| | X | | | X | | | X | |
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AlphaBetaPlayer
        | depth: 1 2 3 4 5 6 7 8 9 10
AlphaBetaAreaPlayer | depth: 1 2 3 4 5 6 7
match: 0 | move: 38 | 1.01s | AlphaBetaAreaPlayer(1) => (11, -2)
| | X | | X | X | | X | |
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 | | X | X | | | | X | X | X | X |
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| | X | | X |
```

```
AlphaBetaAreaPlayer | depth: 1 2 3 4 5 6 7 8
match: 0 \mid move: 40 \mid 1.00s \mid AlphaBetaAreaPlayer(1) \Rightarrow (11, 0)
 1
     - 1
       - 1
        1
| | X | | X | X | | X | | X |
+ - + - + - + - + - + - + - + - + - +
| | X | | | X | | | X | | | X |
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AlphaBetaPlayer
      | depth: 1 2 3 4 5 6 7 8 9 10 11 12
AlphaBetaAreaPlayer | depth: 1 2 3 4 5 6 7 8 9 10
match: 0 | move: 42 | 1.00s | AlphaBetaAreaPlayer(1) => (11, -2)
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          | | X |
1
+ - + - + - + - + - + - + - + - + - +
| | X | | X | X | 1 | X | | X | |
| | X | X | X | | | X | X | | |
| | X |
```

| depth: 1 2 3 4 5 6 7 8 9 10

AlphaBetaPlayer

```
AlphaBetaPlayer
       | depth: 1 2 3 4 5 6 7 8 9 10 11 12
AlphaBetaAreaPlayer | depth: 1 2 3 4 5 6 7 8 9 10
match: 0 | move: 44 | 1.00s | AlphaBetaAreaPlayer(1) => (1, -2)
| 1 |
      1
        + - + - + - + - + - + - + - + - + - +
 | | X | X | X | | | | X | X | | |
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 | X | | | | X | 2 | | X | | |
+ - + - + - + - + - + - + - + - + - +
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AlphaBetaPlayer
       | depth: 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15
AlphaBetaAreaPlayer | depth: 1 2 3 4 5 6 7 8 9 10 11
match: 0 \mid move: 46 \mid 1.00s \mid AlphaBetaAreaPlayer(1) \Rightarrow (2, -1)
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      +-+-+-+-+
```

| | X | X |

| | X | | | | X | X | | X |

| | X | X | X |

AlphaBetaPlayer | depth: 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 AlphaBetaAreaPlayer | depth: 1 2 3 4 5 6 7 8 9 10 11 12 13 14 match: 0 | move: 48 | 1.00s | AlphaBetaAreaPlayer(1) => (2, 1)

+ - + - + - + - + - + - + - + - + - + + - + - + - + - + - + - + - + - + - + 1 1 | | X | | + - + - + - + - + - + - + - + - + - + + - + - + - + - + - + - + - + - + - + | | X | X | X | | | X | X | | | + - + - + - + - + - + - + - + - + - + | | X | 2 | | | X | X | | X | | | +-+-+-+-+

AlphaBetaPlayer | depth: 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19

AlphaBetaAreaPlayer | depth: 1 2 3 4 5 6 7 8 9 10 11 12 13 14 match: 0 | move: 50 | 1.00s | AlphaBetaAreaPlayer(1) => (1, 2)

```
| | X | X | | | X | X | | X | | |
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AlphaBetaPlayer
       | depth: 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19
AlphaBetaAreaPlayer | depth: 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20
21 22 23 24
match: 0 | move: 52 | 0.99s | AlphaBetaAreaPlayer(1) => (1, 2)
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| | | | 1 | X | X | |
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  | | X | | X | X | X | X | X | X |
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| | X | X | | | X | X | | X | | |
AlphaBetaPlayer | depth: 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20
21 22 23 24 25 26 27 -inf
AlphaBetaAreaPlayer | depth: 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20
21 22 23 24 25 26 inf
match: 0 | move: 54 | 0.31s | AlphaBetaAreaPlayer(1) => (12, 1)
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| |2| |1| | X | X | X
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| | X | X | | | X | X | | X | | |
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AlphaBetaPlayer
       | depth: 1 -inf
AlphaBetaAreaPlayer | depth: 1 inf
match: 0 \mid move: 56 \mid 0.00s \mid AlphaBetaAreaPlayer(1) => (11, -2)
| X |
         | X | | X |
| | X | | X | | | X | | X | | X |
+-+-+-+-+
| | 1 | X | 2 | X | X | X | X | X | X | X |
| | X | X | X | | | | X | X | | |
| | X | X | | | X | X | | X |
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+ - + - + - + - + - + - + - + - + - +
AlphaBetaPlayer
       | depth: 1 -inf
AlphaBetaAreaPlayer | depth: 1 inf
match: 0 | move: 58 | 0.00s | AlphaBetaAreaPlayer(1) => (12, 1)
| 1 | | 2 | X | | X | | X |
| | X | | X | | | X | | X | | X | |
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| | X | X | X | | | | X | X | | |
+ - + - + - + - + - + - + - + - + - +
 | X |
AlphaBetaPlayer
       | depth: 1 -inf
AlphaBetaAreaPlayer | depth: 1 inf
match: 0 | move: 60 | 0.00s | AlphaBetaAreaPlayer(1) => (11, -2)
+ - + - + - + - + - + - + - + - + - +
| | X | | X | 1 | | 2 | X | | X | |
| X | X | X | X | X | X | X | X | X | |
+ - + - + - + - + - + - + - + - + - +
        | | X | X |
  | X | X | X |
+ - + - + - + - + - + - + - + - + - +
| X | X | | | X | X |
           | X |
+-+-+-+-+-+-+-+-+
AlphaBetaPlayer
       | depth: 1 -inf
AlphaBetaAreaPlayer | depth: 1 inf
match: 0 | move: 62 | 0.01s | AlphaBetaAreaPlayer(1) => (11, 0)
+-+-+-+-+
| | X | | X | X | | X | X | | X |
```

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+-+-+-+-+
 | | X | X | X | | | | X | X |
| | X | X | | | X | X | | | |
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AlphaBetaPlayer
      | depth: 1 -inf
AlphaBetaAreaPlayer | depth: 1 inf
match: 0 | move: 64 | 0.00s | AlphaBetaAreaPlayer(1) => (12, -3)
+-+-+-+-+
| | X | | X | X | | X | X | 1 | X | |
| | X | X | X | X | X | X | X | X | 2 | X |
| X | X | X | X | X | X | X | X | X | |
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| | X | X | X | | | | X | X |
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AlphaBetaPlayer
      | depth: 1 -inf
match: 0 | move: 65 | 0.00s | AlphaBetaPlayer(0) => (11, 0)
```