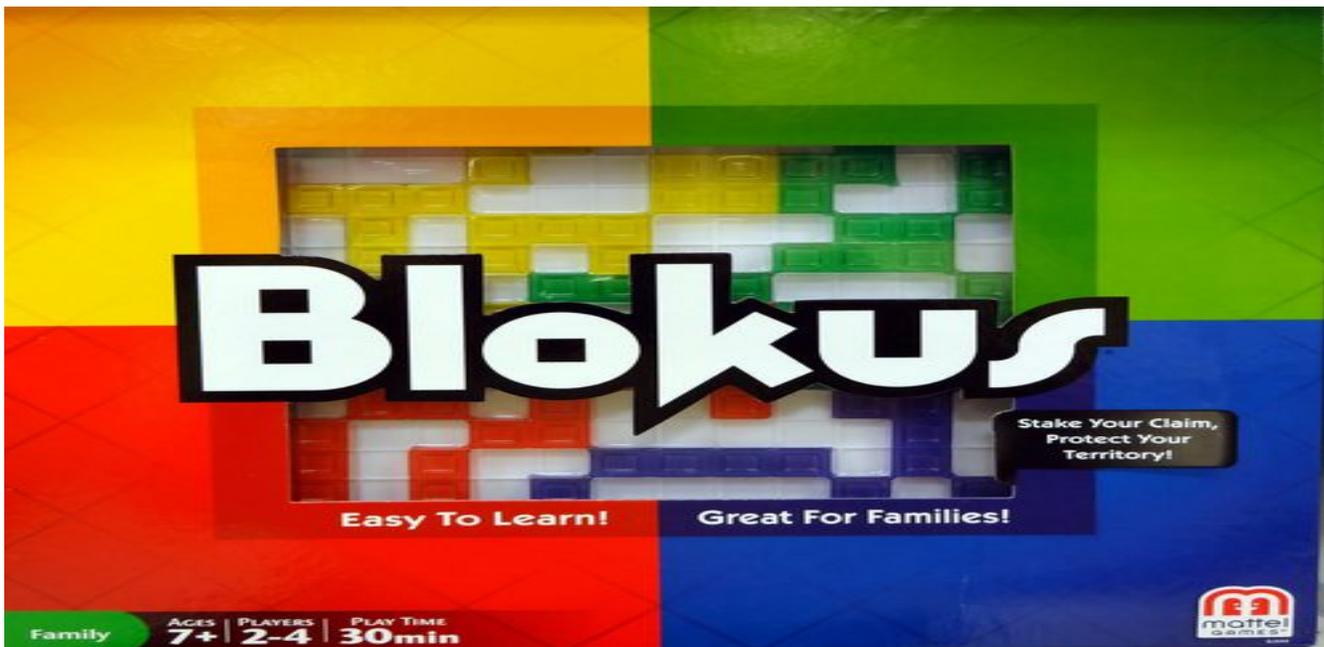


# Blokus Game Solver

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**Figure 1:** Blokus Game Board Cover

## ABSTRACT

Blokus (officially pronounced as “Block us”) is an abstract strategy board game with transparent Tetris-shaped, color pieces that players are trying to play onto the board. However, the players can only place a piece that touches at least one corner of their own pieces on the board. The ultimate goal of the game is to place as many pieces onto the board as a player can while blocking off the opponent’s ability to place more pieces onto the board. Each player has pieces with different shapes and sizes that can be placed onto the board, where each block within a piece counts as one point. The player that scores the highest wins the game.

Just like other strategy board game such as chess, Blokus contains definite strategic patterns that can be solved with computer algorithms. Various algorithms were discovered and created to develop winning strategy and AI

against human opponents. In this work, I am developing random and different greedy strategies to analyze the effectiveness of different factors such as pieces’ size, corner availability, and first-player turn.

## 1 INTRODUCTION

Blokus is an abstract strategy board game like Chess and Chinese checkers. They are board games that can be solved purely with algorithms, which means no randomness or intuition. Every move can be determined based off of various board and game conditions such as current scores and total pieces used. Therefore, AI can be developed to effectively defeat human players in these board game, which allows people to formulate strategies that have a high probability of winning rate [1].

I develop a Blokus game solver to study game algorithms and AI along with exploring a new board game.

Due to abstract strategy board games having solvable patterns, an AI with human behaviors can be developed and evolved as more games are played [1].

There are various known strategies that can be used by the AI to solve abstract strategy board game. The most basic strategy would be simply randomly placing pieces on valid location on the board, which is not an effective strategy since the winning rate is purely dependent on chance. The Greedy algorithm and minimax algorithm are two known strategies to solve these types of strategy games. In Blokus, a basic greedy algorithm in general would make the player put down the largest pieces first before putting the smaller one. This strategy allows the player to get higher scores early on in the game while decrease the number of places on the board the opponent can use. However, there are variations of greedy algorithm that can be implemented in order to improve the effectiveness of the algorithm. By taking account of other variables in Blokus, the effectiveness of the algorithm can be improved.

A more complex decision-making algorithm would be minimax algorithm. It creates a search tree of players making alternate moves with different reward values at the leaf nodes, then the players will choose the child with the best rewards for their cases [2]. Each tree level alternate between different players. However, due to the number of possible moves in Blokus, a complete search tree will hit the memory limitations. Also, the amount of time required to calculate the best moves are computationally expensive. Therefore, the minimax algorithm usually creates a search tree only up to certain number of moves. Techniques such as alpha-beta pruning help decrease the time needed to search the optimal moves by reducing the branching factor in the search tree [3].

Different algorithms represent different strategies that can be used by human players. However, no strategies are perfect, especially in complex strategy game like Blokus. Some strategies do give higher chance of winning but introduce extreme complexity such as the minimax algorithm. While others are simple but not generally effective. Other strategies are effective but contain limitation. Each strategy can also have variations. Monte Carlo tree search is an algorithm derived from the minimax algorithm which can be used for Blokus as well [2].

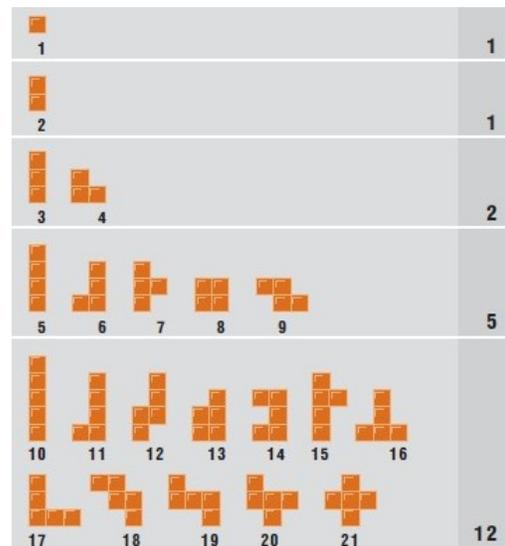
For the algorithm analysis in Blokus, I am focusing on the effectiveness of different algorithms on winning in Blokus. The analysis includes the different factors that can be used in the algorithm when calculating optimal moves and the effectiveness of algorithms in comparison with one another. The players' turn is switched to determine whether a first-player advantage exists in Blokus or not.

## 2 BACKGROUND

Blokus was invented in year 2000 and released by a French company, Sekkoia. It has won many awards such as the 2003 Mensa Select [4]. This game has a deep resemblance to Tetris due to the shapes and blocks of the game pieces despite the big difference in game mechanics and genre. Also, the pieces allow some other various interesting studies to be made beside the actual game play like how squares can be covered on the board without any pieces touching each other [5].

### 2.1 How to Play

Blokus can be play by 2 to 4 people. The board is usually 20 rows by 20 columns; however, two-player version has 14 rows by 14 columns. Each player has 21 difference game pieces that can be place on the game board. The goal of the game for each player is to place as many of his or her own pieces onto the board as possible without violating any placement rule [4].



**Figure 2:** The 21 pieces that each player has at the beginning of the game, which can be rotated and flipped.

The general overview of Blokus gameplay is as follow:

- 1) Each player needs to first place one of his or her piece at one of the corner regions (one square in the piece must touches the corner square)
- 2) Each player takes turn putting down pieces onto the board.
- 3) Each new piece puts down by the player must touch at least one corner of his or her other piece.
- 4) No flat edge should touch one another for each player's pieces (Different player's pieces can touch one another).
- 5) Whenever one of the players is unable to place anymore pieces onto the board, the player must pass his or her turn.

6) The game ends when both players cannot put down anymore pieces onto the board.

7) The score is counted by the number of unit squares in the board for each player.

8) If a player places all of his or her pieces onto the board, then he or she gains an additional 15 points.

9) If a player places all the pieces with the smallest piece for the last one, then he or she gains an additional 5 points.

10) The player with the highest point wins the game.

## 2.2 Blokus Version for Algorithm Study

For this study, the Blokus is based off of the Travel Blokus/Blokus Duo version, which is 2 players with 14 by 14 size board instead of the original 20 by 20 size. The minor modification is the players shall start at the opposite corners instead of the center of the board. This version allows a better strategic study due to no possibility of teaming up, which can happen for 3-4 players. The smaller size also helps demonstrate the effectiveness of different strategies in limited spaces. The limited spaces also mean getting the game bonus for any player is not feasible; therefore, it is not taken account in the algorithm.



**Figure 3:** Blokus Duo game board, the board size the game solver algorithm uses, with the player starting at opposite corners instead of the center.

The algorithm study focuses on the entire game play in general. The player shall mainly focus on maximum piece placements on the board. The two main algorithms in this study are the random and greedy algorithms.

## 2.3 Random Strategy

The player who utilizes random strategy randomly chooses a piece to put onto the board in a valid configuration. The randomness may provide

unpredictability in the player's move. However, it also means the player might place low priority piece first. The algorithm itself is very simple, but it is too unreliable to be an effective agent for Blokus.

## 2.4 Greedy Strategy

The player who utilizes greedy strategy place pieces based on the overall weight of different factors. Some pieces will have higher priority than others depending what is used to calculate their importance. The weight determines how beneficial a piece placement is. The piece with the highest weight will be chosen to be placed. However, if that piece cannot be place in any valid configuration, the next highest one is chosen and so on.

The factors that are used for the algorithm in this study are the size of the piece and the total corner difference between the players after a piece placement.

*2.4.1 Size.* The main focus of the greedy strategy. The main purpose of Blokus is to place as much unit block onto the board as possible. Placing larger size piece first is the fastest way to get higher scores quickly. The wider available spaces also make large size easier to place earlier than later. Having large size pieces later run the risk of them being left over due to no valid placement configuration [6].

*2.4.2 Total Corner Difference Between Players.* The capability to put a new piece onto the board for a player is partly determined by how many corners the players can try to find valid configurations. The more corners, the higher the chance of finding at least one valid configuration. Since the placement of a piece can also cut off the opponent's corners, opponent's corners should be taken into account of as well [6]. If a player has a higher available corner to place pieces than the opponent, the player then has an advantage in terms of possible board placement location and blocking other player's corners.

Due to the size actually matters directly to the score while the corner indirectly affects the possible score, the greedy strategy weighs the size of the piece more than the total corner difference. An advanced greedy strategy involves maximizing the player's own score while keeping its total available corners higher than the opponent. But the calculation of the corner difference increases time complexity of the greedy algorithm overall.

## 2.5 Other Strategies

Greedy strategy is simply one of the sophisticated algorithms to use. Strategies such as the Monte-Carlo and minimax are other alternatives that are effective as well.

Other strategy involves placing specific piece at specific region to cut off the opponents' movement. The Barasona Opening is a starting diagonal board placement and cut off strategy that is meant to split the board to reduce mobility against the opponents. The pieces used are usually very hard to put on the board effectively.

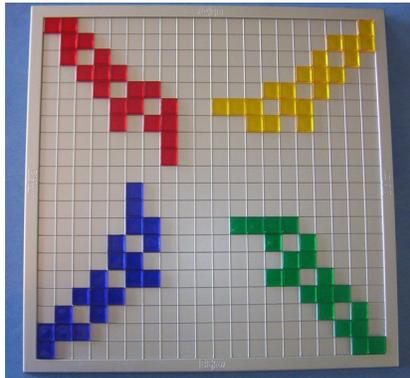


Figure 4: The Barasona Opening Used by All Players

While these other strategies are not used for this algorithm study, they are worth noting for their effectiveness to win in Blokus.

### 3 GAME IMPLEMENTATION

The Blokus simulation is developed in Python 3. The program allows 2 to 4 AIs to play a game of Blokus. The board size can be changed as well. For the purpose of this study, the board size will be 14 by 14 with only two AIs playing the game. The AIs can play Blokus using different strategy that is developed in program. They are random and two variations of the greedy strategies.

#### 3.1 Blokus Pieces

The Blokus pieces are developed under a defined class *Shape* with different id to distinguish different pieces from each other. Each piece contains a list of points to determine to its attempted placement position and a list of corners of the pieces for checking valid placement based on the rule of Blokus. The pieces also support methods that perform the rotation and flipping of different pieces using a reference point on the pieces. A method is defined for each piece to set its points and corners using a reference point:

```
def set_points(self, x, y):
    self.points = [(x, y)]
    self.corners = [(x + 1, y + 1), (x - 1,
        y - 1), (x + 1, y - 1), (x - 1, y + 1)]
```

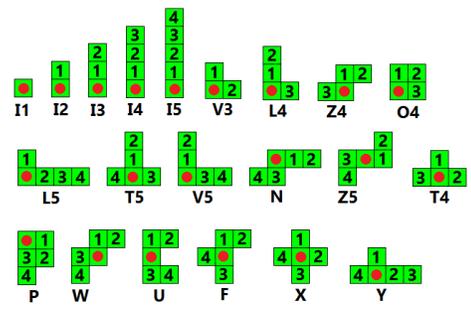


Figure 5: The id of the piece used in the program. The red dot shows an example of the reference point for each piece, which can be changed.

#### 3.2 Board Validation

A *Board* class is defined to keep track of the state of the Blokus board. The object supports capability of updating the board for different players and print the current board configuration for debugging and viewing game progress. Based off from the official rule of the Blokus game [4], several validation methods are implemented to make sure the players are placing new pieces in a valid configuration. Each piece will be checked for adjacent, corner, overlap, and within-bound placement. The validation will be determined by checking each unit square of a piece. The validation will only be successful if all of them pass.

#### 3.3 Blokus Player

A *Player* is defined to represent different Blokus player in a game. Each has a different player id to distinguish from each other. The players keep track of their available game pieces, current corners they have for piece placement, their scores, and the strategy they are using.

Each player shall calculate its own list of possible valid piece placement based on the current states of the board by going through every available piece, every possible rotation and flip on every possible reference point for each piece. The calculation for possible moves required the most computation in the simulation.

Each player's strategy shall determine what method it uses to determine the next move. Those methods will be algorithms to choose the next move.

#### 3.4 Blokus Game

A *Blokus* is defined to represent an instance of a Blokus game. It keeps track of all the players in game, the current round, the game board, and all the initial starting pieces that are given to players. It also facilitates the

simulation of each round of the game, performing all the validation and piece placement of each player.

The *Blokus* class sets up the validation rule for the players to use to determine possible piece placements. An invalid placement shall result in the immediate termination of the game to signify an unexpected error in the simulation that need to be fixed.

The class also determines when the game is finished by checking whether there is any possible move that can be perform by a player in the game. If no player can perform a move, the game is concluded. Then the scores of all players are compared and a winner or a tie is decided.

### 3.4 Player Strategy

Each different strategy is represented by different method that does the necessary calculation to get the next move for the players. In general, all the methods shall use all the player's pieces available piece to get a list of possible piece placements. Then the priority choice of these placements shall be decided by different methods. The random strategy simply chooses a random placement among the list.

The basic greedy strategy simply chooses a piece with the largest size that can be placed onto the board. Since there are multiple pieces that have size of 5 in the game, the strategy chooses randomly among them. The orientation of the piece is random as well.

The advanced greedy strategy sets up a weight for each possible placement using the piece's size and the total corner difference between players after the piece placement. The player chooses the placement with the highest weight. This removes all the randomness in the decision-making process for the piece placement. The weight is defined as:

```
weight = 2 * size + corner difference
corner difference = player's corner - opponent's corner
```

Each strategy in the program is defined as a method. Each player has one of these strategies in its parameters. The random, greedy (basic), and greedy (advanced) are called *Random\_Player*, *Greedy\_Player*, and *Greedy\_Player\_Two* respectively.

## 4 SOLUTION STRATEGY

To study the effectiveness of each game play strategy. I compete two AIs with each other using different or same strategy. For each different study, I make the two AIs play a hundred games to determine the win ratio for different

strategy. The two different greedy strategies are also matched together to determine the effectiveness of different heuristics. In the event of a tie, the result will not count toward a win for either player.

For a match that involves two different strategies, the order of the players is switched to study the impact of a player's first turn on the strategy. This determines whether a first turn advantage exists for Blokus.

The following lists are the matches I set up for the studies (The left side represents the player that goes first in the game):

- Random vs. Random
- Greedy (Basic) vs. Random
- Random vs. Greedy (Basic)
- Greedy (Basic) vs. Greedy (Basic)
- Greedy (Advanced) vs. Random
- Random vs. Greedy (Advanced)
- Greedy (Advanced) vs. Greedy (Advanced)
- Greedy (Advanced) vs. Greedy (Basic)
- Greedy (Basic) vs. Greedy (Advanced)

## 5 GAME RESULTS

Two players with the same strategies can be match together as well. This scenario is treated in terms of 'vs. itself'. Each versus match statistic is in the form of:

**Strategy One vs. Strategy Two: Win, Loss, Tie**

### 5.1 Random Strategy

Two players using random strategy were pitted against each other to evaluate the impact of the first turn advantage. The win ratio of the first turn player is **61, 35, 4**.

### 5.2 Greedy (Basic) Strategy

The match statistic and its turn order for the greedy (basic) player is as follow:

- First Turn vs. Random: **95, 4, 1**
- Second Turn vs. Random: **92, 6, 2**
- First Turn vs. Greedy (Advanced): **64, 32, 4**
- First Turn vs. itself: **59, 38, 3**

### 5.3 Greedy (Advanced) Strategy

The match statistic and its turn order, if applicable, for the greedy (advanced) player is as follow:

- First Turn vs. Random: **86, 13, 1**
- Second Turn vs. Random: **78, 19, 3**

- First Turn vs. Greedy (Basic): **42, 58, 0**
- First Turn vs. itself: **51, 39, 10**

## 6 DISCUSSIONS

### 6.1 Greedy versus Random

The results from the matches show the players using greedy strategies has a clear advantage over the players who place pieces randomly. Both greedy strategies show a winning ratio of above 75% against random strategy no matter if they go first or not.

Since Blokus is a score-based game and placing a larger piece gives a higher score, it provides an incentive to place the largest piece onto the board first. Placing large pieces first has the advantage of getting a higher score early on and taking away a hard piece placement before the chance is lost.

Larger pieces have the general disadvantage of being harder to place on the board due to the space they take up. Once a player lost the chance to put a piece, that player loses many possible points that can be gain. With more than half of the game pieces being large pieces, the loss can quickly stack up. Therefore, putting a larger piece first make sure the players have pieces that are easier to put onto board, even when board is filled up a lot [6]. The greedy algorithm emulates this type of behavior pattern in Blokus.

### 6.2 Greedy Weight Consideration

The results from the matches show the players using greedy (basic) strategy have a higher chance of winning than the players using greedy (advanced) strategy. By factoring the total available corners into the greedy strategy, the effectiveness of the strategy seems to decrease. While it has a distinct advantage over the players using random strategy, it is not effective against a simpler greedy strategy. However, the score difference tends to be closer. The chance of tie also seems to be higher as a result.

The usage of corner consideration in the algorithm makes it more a greedy strategy than the basic version due to each piece placement has a clear weight value that is different from another same size piece or the same piece with different placement configuration.

The greedy (advanced) strategy does not take into consideration the total corner beyond the next move. Therefore, it is possible a player cuts off another player's access to the other side of the board, which drastically reduce a player's possible future placements. This strategy does not take account of the total future possible placements, which would have more impact, since it would help prevent the above scenario from happening often.

The greedy strategy in general does not involve direct attack and defense against the opponent, it only concerns about its own scores. The additional factors besides score in the greedy algorithm would only be counterproductive unless an active attack and defense is implemented as well.

### 6.3 First Turn Advantage

Players in Blokus do experience first turn advantage. When two players using the same strategy match against each other, the first player has a higher chance of winning, which is evident in both stochastic and deterministic strategies. In all three strategies, the win ratio of the first player is higher than the second player. Also, deterministic strategies like greedy (advanced) strategy has a higher tie ratio as well.

Considered the Blokus board has only limited spaces, especially with only 14 by 14 size, players usually will not be able to put every piece onto the board. Each piece put onto the board reduces the spaces even further. Therefore, players who get to put their pieces first have more spaces to work with. They get to set the flows of the game by taking away the opponents' options first.

The decision-making may be random, but no matter what piece any player chooses, the board's available space will always reduce.

### 6.4 Other Considerations

While a general greedy strategy provides a high chance of winning chance in Blokus. There are other specific strategies that can be applied to boost the winning chance even more. The ability to block an opponent's option purposely and put down a complex piece among the same size pieces are keys to gain advantage over other opponents. Certain large pieces such as N, F, W, Y (Pieces ids from *Figure 5*) have various flexible usages that make them crucial in the game [7-8].

The Barasona's opening is an example of reducing the placement capability of other players. If the greedy (advanced) strategy takes the large versatile pieces into considerations, the effectiveness would improve instead of reducing. These strategies would involve more sophisticated algorithm to find the right pieces to use and perform the correct placements.

## 7 CONCLUSIONS

In summary, I have performed an experimental study of the Blokus game play strategy that has been used by actual players. The theory behind various concepts and factors in Blokus is put in practice to see their impacts in an

actual match. The greedy algorithm is shown to be extremely effective in compared to the random algorithm. The simple greedy strategy has a better winning chance than a more complex greedy strategy when no active attack and defense against opponent is involved. The first turn advantage is also a prominent factor in Blokus also, which seems to be a factor that applies to other turn-based abstract strategy board games like Chess and even tiny game like Tic-Tac-Toe.

The development of the Blokus simulation also demonstrates the high time complexity to play multiple games. The different configurations of many board pieces due to the corner availability, rotation, and flip make the algorithms time-consuming. An improvement with the placement search and algorithm needs to be made before an effective AI for Blokus can be developed.

#### ACKNOWLEDGMENTS

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## APPENDIX

### A BLOKUS SIMULATION PROGRAM

The Blokus simulation is developed and ran in Python 3.

#### A.1 shape.py

```

1 import math
2
3 # Get the new x value of point pt (x, y) rotated about reference point
4 # refpt (x, y) by degrees clockwise
5 def rotatex(pt, refpt, deg):
6     return (refpt[0] + (math.cos(math.radians(deg)) * (pt[0] - refpt[0]))
7           + (math.sin(math.radians(deg)) * (pt[1] - refpt[1])))
8
9 # Get the new y value of point pt (x, y) rotated about reference point
10 # refpt (x, y) by degrees clockwise
11 def rotatey(pt, refpt, deg):
12     return (refpt[1] + (-math.sin(math.radians(deg))*(pt[0] - refpt[0]))
13           + (math.cos(math.radians(deg)) * (pt[1] - refpt[1])))
14
15 # Get the new point (x, y) rotated about the reference point refpt (x, y)
16 # by degrees clockwise
17 def rotatep(pt, refpt, deg):
18     return (int(round(rotatex(pt, refpt, deg))),
19           int(round(rotatey(pt, refpt, deg))))
20
21 # The Shape class
22 # Each difference game piece is a subclass of shape
23 # Each has a different id and specific total amount of block (size)
24 # points represent the shape of the pieces
25 # corners represent the corners to the piece
26 class Shape:
27     def __init__(self):
28         self.id = None
29         self.size = 1
30
31     # Set the shapes' point (x, y) locations on the board
32     def set_points(self, x, y):
33         self.points = []
34         self.corners = []
35
36     # Create the shapes on the board, num = square index of the piece
37     # pt = reference point
38     def create(self, num, pt):
39         self.set_points(0, 0)
40         pm = self.points
41         self.pts_map = pm
42
43         self.refpt = pt
44         x = pt[0] - self.pts_map[num][0]
45         y = pt[1] - self.pts_map[num][1]
46         self.set_points(x, y)

```

```

47
48 # Returns the points that would be covered by a
49 # shape that is rotated 0, 90, 180, of 270 degrees
50 # in a clockwise direction.
51 def rotate(self, deg):
52     self.points = [rotatep(pt, self.refpt, deg) for pt in self.points]
53     self.corners = [rotatep(pt, self.refpt, deg) for pt in self.corners]
54
55 # Returns the points that would be covered if the shape
56 # was flipped horizontally or vertically.
57 # orientation = 'h' (horizontal) or 'v' (vertical)
58 # NOTE: For this project, vertical flip isn't needed
59 def flip(self, orientation):
60     # flip horizontally
61     def flip_h(pt):
62         x1 = self.refpt[0]
63         x2 = pt[0]
64         x1 = (x1 - (x2 - x1))
65         return (x1, pt[1])
66
67     # flip the piece horizontally
68     if orientation == 'h':
69         self.points = [flip_h(pt) for pt in self.points]
70         self.corners = [flip_h(pt) for pt in self.corners]
71
72 # List of all the 21 Blokus shape objects
73 class I1(Shape):
74     def __init__(self):
75         self.id = 'I1'
76         self.size = 1
77
78     def set_points(self, x, y):
79         self.points = [(x, y)]
80         self.corners = [(x + 1, y + 1), (x - 1, y - 1), (x + 1, y - 1),
81             (x - 1, y + 1)]
82
83 class I2(Shape):
84     def __init__(self):
85         self.id = 'I2'
86         self.size = 2
87
88     def set_points(self, x, y):
89         self.points = [(x, y), (x, y + 1)]
90         self.corners = [(x - 1, y - 1), (x + 1, y - 1), (x + 1, y + 2),
91             (x - 1, y + 2)]
92
93 class I3(Shape):
94     def __init__(self):
95         self.id = 'I3'
96         self.size = 3
97
98     def set_points(self, x, y):
99         self.points = [(x, y), (x, y + 1), (x, y + 2)]

```

```
100     self.corners = [(x - 1, y - 1), (x + 1, y - 1), (x + 1, y + 3),
101                    (x - 1, y + 3)]
102
103 class I4(Shape):
104     def __init__(self):
105         self.id = 'I4'
106         self.size = 4
107
108     def set_points(self, x, y):
109         self.points = [(x, y), (x, y + 1), (x, y + 2), (x, y + 3)]
110         self.corners = [(x - 1, y - 1), (x + 1, y - 1), (x + 1, y + 4),
111                        (x - 1, y + 4)]
112
113 class I5(Shape):
114     def __init__(self):
115         self.id = 'I5'
116         self.size = 5
117
118     def set_points(self, x, y):
119         self.points = [(x, y), (x, y + 1), (x, y + 2), (x, y + 3), (x, y + 4)]
120         self.corners = [(x - 1, y - 1), (x + 1, y - 1), (x + 1, y + 5),
121                        (x - 1, y + 5)]
122
123 class V3(Shape):
124     def __init__(self):
125         self.id = 'V3'
126         self.size = 3
127
128     def set_points(self, x, y):
129         self.points = [(x, y), (x, y + 1), (x + 1, y)]
130         self.corners = [(x - 1, y - 1), (x + 2, y - 1), (x + 2, y + 1),
131                        (x + 1, y + 2), (x - 1, y + 2)]
132
133 class L4(Shape):
134     def __init__(self):
135         self.id = 'L4'
136         self.size = 4
137
138     def set_points(self, x, y):
139         self.points = [(x, y), (x, y + 1), (x, y + 2), (x + 1, y)]
140         self.corners = [(x - 1, y - 1), (x + 2, y - 1), (x + 2, y + 1),
141                        (x + 1, y + 3), (x - 1, y + 3)]
142
143 class Z4(Shape):
144     def __init__(self):
145         self.id = 'Z4'
146         self.size = 4
147
148     def set_points(self, x, y):
149         self.points = [(x, y), (x, y + 1), (x + 1, y + 1), (x - 1, y)]
150         self.corners = [(x - 2, y - 1), (x + 1, y - 1), (x + 2, y),
151                        (x + 2, y + 2), (x - 1, y + 2), (x - 2, y + 1)]
152
```

```
153 class O4(Shape):
154     def __init__(self):
155         self.id = 'O4'
156         self.size = 4
157
158     def set_points(self, x, y):
159         self.points = [(x, y), (x, y + 1), (x + 1, y + 1), (x + 1, y)]
160         self.corners = [(x - 1, y - 1), (x + 2, y - 1), (x + 2, y + 2),
161             (x - 1, y + 2)]
162
163 class L5(Shape):
164     def __init__(self):
165         self.id = 'L5'
166         self.size = 5
167
168     def set_points(self, x, y):
169         self.points = [(x, y), (x, y + 1), (x + 1, y), (x + 2, y), (x + 3, y)]
170         self.corners = [(x - 1, y - 1), (x + 4, y - 1), (x + 4, y + 1),
171             (x + 1, y + 2), (x - 1, y + 2)]
172
173 class T5(Shape):
174     def __init__(self):
175         self.id = 'T5'
176         self.size = 5
177
178     def set_points(self, x, y):
179         self.points = [(x, y), (x, y + 1), (x, y + 2), (x - 1, y), (x + 1, y)]
180         self.corners = [(x + 2, y - 1), (x + 2, y + 1), (x + 1, y + 3),
181             (x - 1, y + 3), (x - 2, y + 1), (x - 2, y - 1)]
182
183 class V5(Shape):
184     def __init__(self):
185         self.id = 'V5'
186         self.size = 5
187
188     def set_points(self, x, y):
189         self.points = [(x, y), (x, y + 1), (x, y + 2), (x + 1, y), (x + 2, y)]
190         self.corners = [(x - 1, y - 1), (x + 3, y - 1), (x + 3, y + 1),
191             (x + 1, y + 3), (x - 1, y + 3)]
192
193 class N(Shape):
194     def __init__(self):
195         self.id = 'N'
196         self.size = 5
197
198     def set_points(self, x, y):
199         self.points = [(x, y), (x + 1, y), (x + 2, y), (x, y - 1), (x - 1, y - 1)]
200         self.corners = [(x + 1, y - 2), (x + 3, y - 1), (x + 3, y + 1),
201             (x - 1, y + 1), (x - 2, y), (x - 2, y - 2)]
202
203 class Z5(Shape):
204     def __init__(self):
205         self.id = 'Z5'
```

```
206     self.size = 5
207
208     def set_points(self, x, y):
209         self.points = [(x, y), (x + 1, y), (x + 1, y + 1), (x - 1, y),
210                        (x - 1, y - 1)]
211         self.corners = [(x + 2, y - 1), (x + 2, y + 2), (x, y + 2),
212                        (x - 2, y + 1), (x - 2, y - 2), (x, y - 2)]
213
214 class T4(Shape):
215     def __init__(self):
216         self.id = 'T4'
217         self.size = 4
218
219     def set_points(self, x, y):
220         self.points = [(x, y), (x, y + 1), (x + 1, y), (x - 1, y)]
221         self.corners = [(x + 2, y - 1), (x + 2, y + 1), (x + 1, y + 2),
222                        (x - 1, y + 2), (x - 2, y + 1), (x - 2, y - 1)]
223
224 class P(Shape):
225     def __init__(self):
226         self.id = 'P'
227         self.size = 5
228
229     def set_points(self, x, y):
230         self.points = [(x, y), (x + 1, y), (x + 1, y - 1), (x, y - 1), (x, y - 2)]
231         self.corners = [(x + 1, y - 3), (x + 2, y - 2), (x + 2, y + 1),
232                        (x - 1, y + 1), (x - 1, y - 3)]
233
234 class W(Shape):
235     def __init__(self):
236         self.id = 'W'
237         self.size = 5
238
239     def set_points(self, x, y):
240         self.points = [(x, y), (x, y + 1), (x + 1, y + 1), (x - 1, y),
241                        (x - 1, y - 1)]
242         self.corners = [(x + 1, y - 1), (x + 2, y), (x + 2, y + 2),
243                        (x - 1, y + 2), (x - 2, y + 1), (x - 2, y - 2), (x, y - 2)]
244
245 class U(Shape):
246     def __init__(self):
247         self.id = 'U'
248         self.size = 5
249
250     def set_points(self, x, y):
251         self.points = [(x, y), (x, y + 1), (x + 1, y + 1), (x, y - 1),
252                        (x + 1, y - 1)]
253         self.corners = [(x + 2, y - 2), (x + 2, y), (x + 2, y + 2),
254                        (x - 1, y + 2), (x - 1, y - 2)]
255
256 class F(Shape):
257     def __init__(self):
258         self.id = 'F'
```

```

259     self.size = 5
260
261     def set_points(self, x, y):
262         self.points = [(x, y), (x, y + 1), (x + 1, y + 1), (x, y - 1), (x - 1, y)]
263         self.corners = [(x + 1, y - 2), (x + 2, y), (x + 2, y + 2),
264             (x - 1, y + 2), (x - 2, y + 1), (x - 2, y - 1), (x - 1, y - 2)]
265
266 class X(Shape):
267     def __init__(self):
268         self.id = 'X'
269         self.size = 5
270
271     def set_points(self, x, y):
272         self.points = [(x, y), (x, y + 1), (x + 1, y), (x, y - 1), (x - 1, y)]
273         self.corners = [(x + 1, y - 2), (x + 2, y - 1), (x + 2, y + 1),
274             (x + 1, y + 2), (x - 1, y + 2), (x - 2, y + 1), (x - 2, y - 1),
275             (x - 1, y - 2)]
276
277 class Y(Shape):
278     def __init__(self):
279         self.id = 'Y'
280         self.size = 5
281
282     def set_points(self, x, y):
283         self.points = [(x, y), (x, y + 1), (x + 1, y), (x + 2, y), (x - 1, y)]
284         self.corners = [(x + 3, y - 1), (x + 3, y + 1), (x + 1, y + 2),
285             (x - 1, y + 2), (x - 2, y + 1), (x - 2, y - 1)]

```

## A.2 blokus.py

```

1 import sys
2 import math
3 import random
4 import copy
5 import shape
6
7 # Blokus Board
8 class Board:
9     # '_' represents empty square
10    # board size: (row: nrow, col: ncol)
11    def __init__(self, nrow, ncol):
12        self.nrow = nrow # total rows
13        self.ncol = ncol # total columns
14        self.state = [['_'] * ncol for i in range(nrow)] # empty board
15
16    # Takes in a player id and a move as a
17    # list of position (x, y) that represent the piece location.
18    def update(self, player_id, placement):
19        for row in range(self.nrow):
20            for col in range(self.ncol):
21                if (col, row) in placement:
22                    self.state[row][col] = player_id
23

```

```

24 # Check if the point (y, x) is within the board's bound
25 def in_bounds(self, point):
26     return 0 <= point[0] < self.ncol and 0 <= point[1] < self.nrow
27
28 # Check if a piece placement overlap another piece on the board
29 def overlap(self, placement):
30     return False in [(self.state[y][x] == '_') for x, y in placement]
31
32 # Checks if a piece placement is adjacent to any square on
33 # the board which are occupied by the player proposing the move.
34 def adj(self, player_id, placement):
35     adjacents = []
36
37     # Check left, right, up, down for adjacent square
38     for x, y in placement:
39         if self.in_bounds((x + 1, y)):
40             adjacents += [self.state[y][x + 1] == player_id]
41         if self.in_bounds((x - 1, y)):
42             adjacents += [self.state[y][x - 1] == player_id]
43         if self.in_bounds((x, y - 1)):
44             adjacents += [self.state[y - 1][x] == player_id]
45         if self.in_bounds((x, y + 1)):
46             adjacents += [self.state[y + 1][x] == player_id]
47
48     return True in adjacents
49
50 # Check if a piece placement is cornering
51 # any pieces of the player proposing the move.
52 def corner(self, player_id, placement):
53     corners = []
54
55     # check the corner square from the placement
56     for x, y in placement:
57         if self.in_bounds((x + 1, y + 1)):
58             corners += [self.state[y + 1][x + 1] == player_id]
59         if self.in_bounds((x - 1, y - 1)):
60             corners += [self.state[y - 1][x - 1] == player_id]
61         if self.in_bounds((x + 1, y - 1)):
62             corners += [self.state[y - 1][x + 1] == player_id]
63         if self.in_bounds((x - 1, y + 1)):
64             corners += [self.state[y + 1][x - 1] == player_id]
65
66     return True in corners
67
68 # Print the current board layout
69 def print_board(self):
70     print("Current Board Layout:")
71     for row in range(len(self.state)):
72         for col in range(len(self.state[0])):
73             print(" " + str(self.state[row][col]), end = ' ')
74         print()
75
76 # Player Class

```

```

77 class Player:
78     def __init__(self, id, strategy):
79         self.id = id # player's id
80         self.pieces = [] # player's unused game piece, list of Shape
81         self.corners = set() # current valid corners on board
82         self.strategy = strategy # player's strategy
83         self.score = 0 # player's current score
84
85     # Add the player's initial pieces for a game
86     def add_pieces(self, pieces):
87         random.shuffle(pieces)
88         self.pieces = pieces
89
90     # Remove a player's piece (Shape)
91     def remove_piece(self, piece):
92         self.pieces = [p for p in self.pieces if p.id != piece.id]
93
94     # Set the available starting corners for players
95     def start_corner(self, p):
96         self.corners = set([p])
97
98     # Updates player information after placing a board piece (Shape)
99     # like the player's score
100    def update_player(self, piece, board):
101        self.score += piece.size # update score
102        if len(self.pieces) == 1: # If the current piece is the last unused piece
103            self.score += 15 # bonus for putting all pieces
104            if piece.id == 'I1':
105                self.score += 5 # bonus for putting the smallest piece last
106            for c in piece.corners: # Add the player's available corners
107                if board.in_bounds(c) and not board.overlap([c]):
108                    self.corners.add(c)
109
110    # Get a unique list of all possible placements (Shape)
111    # on the board
112    def possible_moves(self, pieces, game):
113        # Updates the corners of the player, in case the
114        # corners have been covered by another player's pieces.
115        self.corners = set([(x, y) for (x, y) in self.corners
116                            if game.board.state[y][x] == '_'])
117
118        placements = [] # a list of possible placements (Shape)
119        visited = [] # a list placements (a set of points on board)
120
121        # Check every available corners
122        for cr in self.corners:
123            # Check every available pieces
124            for sh in pieces:
125                # Check every reference point the piece could have.
126                for num in range(sh.size):
127                    # Check every flip
128                    for flip in ["h", "v"]:
129                        # Check every rotation

```

```

130         for rot in [0, 90, 180, 270]:
131             # Create a copy to prevent an overwrite on the original
132             candidate = copy.deepcopy(sh)
133             candidate.create(num, cr)
134             candidate.flip(flip)
135             candidate.rotate(rot)
136             # If the placement is valid and new
137             if game.valid_move(self, candidate.points):
138                 if not set(candidate.points) in visited:
139                     placements.append(candidate)
140                     visited.append(set(candidate.points))
141         return placements
142
143     # Get the next move based off of the player's strategy
144     def next_move(self, game):
145         return self.strategy(self, game)
146
147 # Blokus Game class
148 class Blokus:
149     def __init__(self, players, board, all_pieces):
150         self.players = players # list of players in the game
151         self.rounds = 0 # current round in the game
152         self.board = board # the game's board
153         self.all_pieces = all_pieces # all the initial pieces in the game
154         self.previous = 0 # previous total available moves from all players
155         self.repeat = 0 # counter for how many times the total available moves are
156                         # the same by checking previous round
157         self.win_player = 0 # winner
158
159     # Check for the winner (or tied) in the game and return the winner's id.
160     # Or return nothing if the game can still progress
161     def winner(self):
162         # get all possible moves for all players
163         moves = [p.possible_moves(p.pieces, self) for p in self.players]
164
165         # check how many rounds the total available moves from all players
166         # are the same and increment the counter if so
167         if self.previous == sum([len(mv) for mv in moves]):
168             self.repeat += 1
169         else:
170             self.repeat = 0
171
172         # if there is still moves possible or total available moves remain
173         # static for too many rounds (repeat reaches over a certain threshold)
174         if False in [len(mv) == 0 for mv in moves] and self.repeat < 4:
175             self.previous = sum([len(mv) for mv in moves])
176             return None # Nothing to return to continue the game
177         else: # No more move available, the game ends
178             # order the players by highest score first
179             candidates = [(p.score, p.id) for p in self.players]
180             candidates.sort(key = lambda x: x[0], reverse = True)
181             highest = candidates[0][0]
182             result = [candidates[0][1]]

```

```

183         for candidate in candidates[1:]: # check for tied score
184             if highest == candidate[0]:
185                 result += [candidate[1]]
186         return result # get all the highest score players
187
188 # Check if a player's move is valid, including board bounds, pieces' overlap,
189 # adjacency, and corners.
190 def valid_move(self, player, placement):
191     if self.rounds < len(self.players): # Check for starting corner
192         return not ((False in [self.board.in_bounds(pt) for pt in placement])
193                    or self.board.overlap(placement)
194                    or not (True in [(pt in player.corners) for pt in placement]))
195     return not ((False in [self.board.in_bounds(pt) for pt in placement])
196                or self.board.overlap(placement)
197                or self.board.adj(player.id, placement)
198                or not self.board.corner(player.id, placement))
199
200 # Play the game with the list of player sequentially until the
201 # game ended (no more pieces can be placed for any player)
202 def play(self):
203     # At the beginning of the game, it should
204     # give the players their pieces and a corner to start.
205     if self.rounds == 0: # set up starting corners and players' initial pieces
206         max_x = self.board.ncol - 1
207         max_y = self.board.nrow - 1
208         starts = [(0, 0), (max_x, max_y), (0, max_y), (max_x, 0)]
209
210         for i in range(len(self.players)):
211             self.players[i].add_pieces(list(self.all_pieces))
212             self.players[i].start_corner(starts[i])
213
214     winner = self.winner() # get game status
215     if winner is None: # no winner, the game continues
216         current = self.players[0] # get current player
217         proposal = current.next_move(self) # get the next move based on
218                                     # the player's strategy
219     if proposal is not None: # if there a possible proposed move
220         # check if the move is valid
221         if self.valid_move(current, proposal.points):
222             # update the board and the player status
223             self.board.update(current.id, proposal.points)
224             current.update_player(proposal, self.board)
225             current.remove_piece(proposal) # remove used piece
226         else: # end the game if an invalid move is proposed
227             raise Exception("Invalid move by player " + str(current.id))
228         # put the current player to the back of the queue
229         first = self.players.pop(0)
230         self.players += [first]
231         self.rounds += 1 # update game round
232     else: # a winner (or tied) is found
233         if len(winner) == 1: # if the game results in a winner
234             self.win_player = winner[0]
235             print('Game over! The winner is: ' + str(winner[0]))

```

```

236         else: # if the game results in a tie
237             print('Game over! Tied between players: '
238                   + ', '.join(map(str, winner)))
239
240 # Random Strategy: choose an available piece randomly
241 def Random_Player(player, game):
242     options = [p for p in player.pieces] # get all player's available pieces
243     while len(options) > 0: # if there are still options to find possible moves
244         piece = random.choice(options) # get a random piece
245         # get a list of all possible moves from that piece
246         possibles = player.possible_moves([piece], game)
247
248         if len(possibles) != 0: # if there is possible moves
249             return random.choice(possibles) # choose a random placements to use
250         else: # no possible move for that piece
251             options.remove(piece) # remove it from the options
252     return None # no possible move left
253
254 # Basic Greedy Strategy: chooses an available piece with the highest size
255 def Greedy_Player(player, game):
256     options = [p for p in player.pieces]
257     # order the piece based on highest size first
258     options.sort(reverse = True, key = lambda x: x.size)
259
260     while len(options) > 0:
261         piece = options[0] # get the largest piece
262         possibles = player.possible_moves([piece], game)
263
264         if len(possibles) != 0:
265             return random.choice(possibles)
266         else:
267             options.remove(piece)
268     return None
269
270 # Advanced Greedy Strategy: chooses an available piece based on a hueristic
271 # It is based on the piece's size and the total corner difference from
272 # its placement
273 def Greedy_Player_Two(player, game):
274     shape_options = [p for p in player.pieces]
275     board = game.board
276     weights = [] # array of tuples, (piece's placement, weight)
277
278     for piece in shape_options:
279         possibles = player.possible_moves([piece], game)
280         if len(possibles) != 0:
281             for possible in possibles:
282                 # set a test player and board to simulate a future move,
283                 # then determine the average total available corners difference
284                 # between the player and its opponents
285                 test_players = copy.deepcopy(game.players)
286                 opponents = [p for p in test_players if p.id != player.id]
287                 test_board = copy.deepcopy(board)
288                 test_board.update(player.id, possible.points)

```

```

289         test_player = copy.deepcopy(player)
290         test_player.update_player(possible, test_board)
291         my_corners = len(test_player.corners)
292         total = 0 # total corner difference between player and each opponent
293         for opponent in opponents:
294             opponent.corners = set([(x, y) for (x, y) in opponent.corners
295                                     if test_board.state[y][x] == '_'])
296             total += (my_corners - len(opponent.corners))
297             average = total / len(opponents) # average corner difference
298             weights += [(possible, 2 * piece.size + average)]
299         weights.sort(key = lambda x: x[1], reverse = True) # sort by highest weight
300         if len(weights) != 0:
301             return weights[0][0] # get the highest weighted placement
302         else:
303             return None # no possible move left
304
305 # Play a game of blokus without showing the board
306 def test_blokus(blokus):
307     blokus.play()
308     # game continues until a winner (or tied) is decided
309     while blokus.winner() is None:
310         blokus.play()
311
312 # play a round of blokus including printing the board
313 def play_blokus(blokus):
314     print("Round: " + str(blokus.rounds))
315     blokus.board.print_board()
316     print('=====')
317     blokus.play()
318     print("Round: " + str(blokus.rounds))
319     blokus.board.print_board()
320     for player in blokus.players:
321         print("Player " + str(player.id) + " score " + str(player.score) + ": "
322               + str([sh.id for sh in player.pieces]))
323     print('=====')
324
325     while blokus.winner() is None:
326         blokus.play()
327         print("Round: " + str(blokus.rounds))
328         blokus.board.print_board()
329         for player in blokus.players:
330             print("Player " + str(player.id) + " score " + str(player.score) + ": "
331                   + str([sh.id for sh in player.pieces]))
332         print('=====')
333
334 # run multiple blokus games with a strategy for each player
335 # Precondition: Only two players
336 def multi_run(printout, repeat, one, two):
337     winner = {1 : 0, 2 : 0} # player one and two's scores
338     for i in range(repeat): # Play multiple times
339         print("New Game " + str(i))
340         order = []
341         first = Player(1, one) # first player

```

```
342     second = Player(2, two) # second player
343     all_pieces = [shape.I1(), shape.I2(), shape.I3(), shape.I4(), shape.I5(),
344                  shape.V3(), shape.L4(), shape.Z4(), shape.O4(), shape.L5(),
345                  shape.T5(), shape.V5(), shape.N(), shape.Z5(), shape.T4(),
346                  shape.P(), shape.W(), shape.U(), shape.F(), shape.X(),
347                  shape.Y()] # set up all the initial game pieces
348     board = Board(14, 14) # 14 by 14 board
349     order = [first, second] # order of the player in the game
350     blokus = Blokus(order, board, all_pieces)
351     if printout: # print or not print the board each round
352         play_blokus(blokus)
353     else:
354         test_blokus(blokus)
355
356     blokus.board.print_board() # print the final board
357     blokus.play()
358     print("Final Score:")
359     plist = sorted(blokus.players, key = lambda p: p.id)
360
361     for player in plist:
362         print("Player " + str(player.id) + ": " + str(player.score))
363     if blokus.win_player > 0: # if there is a winner, not a tie
364         winner[blokus.win_player] += 1 # update the winner's win count
365     # print players' win count
366     print("Player one win count: " + str(winner[1]))
367     print("Player two win count: " + str(winner[2]))
368     print()
369
370
371 def main():
372     printout = False
373     if len(sys.argv[1:]) > 0 and sys.argv[1] == 'print':
374         printout = True
375     print("Senior Project Blokus Game")
376
377     # For the project, play each competition for 100 games
378     # Three possible strategies in the game:
379     # Random_Player, Greedy_Player, Greedy_Player_Two
380     multi_run(printout, 100, Random_Player, Greedy_Player_Two)
381
382
383 if __name__ == '__main__':
384     main()
```