

# Project 1: Gomoku

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## 1. Preliminaries

This project is doing some simple implementations on Gomoku. Gomoku, also called Five in a Row, is an abstract strategy board game. It is traditionally played with Go pieces (black and white stones) on a Go board, using 15x15 of the 19x19 grid intersection. [1] Gomoku has less complicity than Go, which has encouraged Deep Mind to develop powerful AI engine AlphaGo in 2016. So Gomoku will be a better particle for beginner. Design of Gomoku AI in this project try to follow the minMax tree with alpha-beta punching, which powered AI Deep Blue to challenge top chess player successfully in the first time.

### 1.1. Software

This project is written by Python 3.7 with editor Atom and Vim. Numpy library is being used.

### 1.2. Algorithm

Depth-first search in the form of minMax tree with alpha-beta punching is used. The method being used is recursion. I defined three group of functions in order to find out next chess position from input chess board. Including given functions, here are nine functions in go.py, and one is for testing.

## 2. Methodology

Different chess layout present different chance to win. For example, live-four and double-three must lead to success. So I evaluate the score of possible layout in final level of DFS. For every layout, I scan each line in vertical, horizontal and diagonal to calculate score and sum three together valued as whole chess layout score. Then, all score were judged by minMax tree to found out best choose position. To speed up minMax tree, alpha-beta punching was used to reduce the number of node to expand.

### 2.1. Representation

Some main data are maintain during process: color, candidate list, chessboard. Others data would be specified inside functions.

- **color**: The chess color of current AI
- **candidate\_list**: Candidate point coordination, nested list. Final tuple in list would be return as final choice.
- **chessboard**: The chessboard layout from player.
- **depth**: The search depth

### 2.2. Architecture

Here list all functions in given class AI:

- Given function:
  - **\_\_init\_\_**: initial function, including global variables.
  - **first\_chess**: if AI is black chess, this function run.
  - **go**: receive current chessboard layout, and return AI's choice.
- Self-defined function:
  - **alphaBeta**: minMax tree and alpha-beta punching.
  - **genNext**: generate available position on chessboard.
  - **evaluateState**: evaluate current chessboard score of black or white
  - **evaluateLine**: evaluate current line score of black or white
  - **mapValue**: map different chess layout in single line to score

The class AI would be executed in test platform.

### 2.3. Detail of Algorithm

Here describes some vital functions.

- **alphaBeta**: use minMax tree with alpha-beta punching to choose best position

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**Algorithm 1** alphaBeta

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**Input:** *source\_chessboard, depth, alpha, beta, chess\_color***Output:** *max\_value*

```
1: if (depth == 0) then
2:
3:   return evaluateState(source_chessboard)
4: end if
5: child_chessboard ← source_chessboard
6: candidate_points ← genNext(child_chessboard)
7: max_value ← alpha
8: max_point ← candidate_points[0]
9: for point in candidate_points do
10:  child_chessboard[point] ← chess_color {action}
11:  value ← -alphaBeta(child_chessboard, depth -
12:    1, -beta, -alpha, -chess_color) {recursion}
13:  child_chessboard[point] ← 0 {undo action}
14:  if (value > max_value) then
15:    max_value ← value
16:    max_point ← point
17:    if (max_value >= beta) then
18:      break
19:    end if
20:  end for
21: if (depth == initial_depth) then
22:  candidate_list.append(max_point) {add max
23:    value point to candidate list}
24: end if
25: return max_value
```

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- **genNext:** generate most possible position

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**Algorithm 2** genNext

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**Input:** *chess\_chessboard, chess\_color***Output:** *score*

```
pos ← emptyposition
2: neighbor_points ← [] {initial}
for point in pos do
4:  if (point has any neighbor in 4 direction) then
5:    add point to neighbor_points
6:  break
7: end if
8: end for
return neighbor_points
```

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- **evaluateState:** return current chessboard layout score

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**Algorithm 3** evaluateState

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**Input:** *chessboard, chess\_color***Output:** *score*

```
score ← 0 {Initial score}
for every line in horizontal, vertical, diagonal do
3:  score ← score + evaluateLine(line, chess_color)
end for
return score
```

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- **evaluateLine:** get single line score

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**Algorithm 4** evaluateLine

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**Input:** *chessline, chess\_color***Output:** *score*

```
score ← 0
continue_point ← 1
block_point ← 0
4: while point in line do
5:   {found own chess} if point == chess_color then
6:     continue_point ← 1
7:     block_point ← 0
8:   if (point == -chess_color) then
9:     block_point ← block_point + 1
10:  end if
11:  point ← next point in line
12:  while point still in line and point ==
13:    chess_color do
14:    point ← next point in line
15:    continue_point ← continue_point + 1
16:    if point == -chess_color then
17:      block_point ← block_point + 1
18:    end if
19:  end while
20:  score ← score +
21:    mapValue(continue_point, block_point)
22: end if
23: point ← next point in line
24: end while
return score
```

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### 3. Empirical Verification

Empirical verification is compared with given test\_code.py and public test platform.

#### 3.1. Design

Successfully to evaluate chessboard score currently. When depth is two, program output current answer within five second. When I try to run in depth of four, right answer almost take half of hour. In a word, more powerful evaluate function and punching(heuristic function) are required.

#### 3.2. Data and data structure

I use test data provided by code\_test.py to test. Numpy matrix are used widely.

#### 3.3. Performance

Right output in depth of two and time of five second. It run so slow when depth is four.

#### 3.4. Result

Pass all test but fail to attend completion.

### **3.5. Analysis**

The evaluate chessboard layout score cost linear time. minMax tree cost most time. The deeper, the more time it cost.

### **Acknowledgment**

Thanks TA Yao Zhao who explain question and provide general method to solve it. And I also thanks for Kebin Sun help me debug my algorithm mistake.

### **References**

- [1] Wikipedia contributors, [Online], Available: <https://en.wikipedia.org/wiki/Gomoku>, [Accessed: 31- Oct- 2018].