

# The Gomoku AI

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## 1. Preliminaries

This is a project of Artificial Intelligence, which aims at implementing a simple AI of Gomoku. Gomoku, also called five in a row, is a board game for two players, in which the aim is to place five pieces with the same color in a row without block. As a game with complete information, people obviously want to calculate all the steps to guarantee to win. Considering the huge calculation that needs to be dealt with, even predicting 5 steps is very hard. In this project, I choose to implement a one-step algorithm.

### 1.1. Software

This project is implemented in Python using IDE PyCharm. The libraries being used include Numpy, Re and Random.

### 1.2. Algorithm

The algorithm used in this project is mainly heuristic search. The main method to implement this search algorithm and some other functions is iteration. There totally have five functions in this project's Python code file.

## 2. Methodology

The rule of Gomoku is very simple, you just need to try your best to place five stones in the same row and avoid your opponent winning ahead of you. This part will show how I implement the main structure of the Gomoku AI.

### 2.1. Representation

First of all, it is unavoidable to consider how to represent the data you use in playing gomoku, such

as black stone, white stone, the chessboard, the choice your program makes at last and so on. The following list shows some data structures I use to represent data and the chess patterns I use.

- **Chessboard:** a numpy two-dimensional array.
- **List of candidate position:** list of tuple.
- **Output position:** the last tuple in the candidate list.
- **Chess color:** the color of stone:
  - color\_white: use 1 to represent
  - color\_black: use -1 to represent
- **Chess patterns[1]:**
  - five: 300000 pts
  - live\_four: 100000 pts
  - cut\_four: 30000 pts
  - live\_three: 15000 pts
  - cut\_three: 5000 pts
  - live\_two: 3000 pts
  - cut\_two: 1000 pts

### 2.2. Architecture

Here are all functions in Python file gobang with pseudocode format.

- **Given:**
  - \_\_initial\_\_: initial the AI.
  - go: read a chessboard and output the decision.
- **Self define:**
  - get\_candidate\_simple: find all the valid position.
  - check\_neighbors: check whether in a 4\*4 field exists a stone.
  - get\_score: count the score if AI places stone here.
  - get\_stone\_around: count stones of mine and opponent's.

## 2.3. Detail of Algorithm

Here show the pseudocode of all the functions.

- **check\_neighbors:** Use transversal to search the chessboard, if there is no stone around this position, it is useless to place a stone here

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**Algorithm 1** check whether in a 4\*4 field exists a stone.

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**Input:** *tuple* index, *np.array* chessboard

**Output:** *true or false*

```
1: function CHECK_NEIGHBORS(chessboard, index)
2:  $x, y \leftarrow \text{index}[0], \text{index}[1]$ 
3:   for  $i = -3 \rightarrow 3$  do
4:     for  $j = -3 \rightarrow 3$  do
5:       if  $i == 0 \& j == 0$  then
6:         continue
7:       else if  $0 \leq x + i \leq 14 \& 0 \leq y + j \leq 14 \& (x + i, y + j)$  is empty then
8:         return true
9:       end if
10:    end for
11:  end for
12:  return false
13: end function
```

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- **get\_candidate\_simple:** Search the whole chessboard to find all valid position

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**Algorithm 2** find all the valid position.

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**Input:** *np.array* chessboard

**Output:** *list* candidatelist

```
1: function GET_CANDIDATE_SIMPLE(chessboard)
2:  $\text{list} \leftarrow []$ 
3:   for  $i = 0 \rightarrow 14$  do
4:     for  $j = 0 \rightarrow 14$  do
5:       if chessboard[i][j] is empty
        & check_neighbors(chessboard, (i, j)) then
6:          $\text{list.add}[i, j]$ 
7:       end if
8:     end for
9:   end for
10:  return candidatelist
11: end function
```

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- **get\_score:** Check all the chess pattern if place a stone here to measure the importance of this position

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**Algorithm 3** calculate the score of chess pattern.

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**Input:** *np.array* chessboard, *tuple* index, *int* color

**Output:** *int* score

```
1: function GET_SCORE(chessboard, index, color)
2:  $x, y \leftarrow \text{index}[0], \text{index}[1]$ 
3:  $\text{score} \leftarrow 0$ 
4:  $\text{pattern\_list} \leftarrow \{ \}$ 
5:  $\text{position} \leftarrow \{\text{left\_to\_right}, \text{up\_to\_down},$ 
    $\text{leftdown\_to\_rightup}, \text{leftup\_to\_rightdown}\}$ 
6:   for  $i$  in  $\text{position}$  do
7:     find chess pattern  $p$  by searching  $i$  at  $(x, y)$ 
8:     if  $p$  in Chess patterns then
9:        $\text{score} += p's \text{ score}$ 
10:       $\text{pattern\_list}[p] += 1$ 
11:    // record the number and type of pattern
12:    end if
13:  end for
14:  return score, pattern_list
15: end function
```

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- **get\_stone\_around:** Search how many stone around here. more concentrated the stones are located, more likely to get advantage

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**Algorithm 4** count stones of mine and opponent's.

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**Input:** *np.array* chessboard, *tuple* index, *int* color

**Output:** *enemy, mine*

```
1: function GET_STONE_AROUND(chessboard, index, color)
2:  $x, y \leftarrow \text{index}[0], \text{index}[1]$ 
3:  $\text{enemy}, \text{mine} \leftarrow 0$ 
4:   for  $i = -1 \rightarrow 1$  do
5:     for  $j = -1 \rightarrow 1$  do
6:       if  $i == 0 \& j == 0$  then
7:         continue
8:       else if  $14 \geq x + i \geq 0 \& 14 \geq y + j \geq 0$  then
9:         if chessboard[x + i][y + j] == -color then
10:            $\text{enemy} += 1$ 
11:         else if
12:           then  $\text{mine} += 1$ 
13:         end if
14:       end if
15:     end for
16:   end for
17:  return enemy, mine
18: end function
```

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- **go**: Given a chessboard, search the whole board to determine which position will get the largest benefit.

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**Algorithm 5** output the final decision position.

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**Input:** *np.array* chessboard

**Output:** the last tuple in candidatelist

```

1: function GO(chessboard)
2: list  $\leftarrow$  get_candidate_simple(chessboard)
3: maxscore  $\leftarrow$  -1 candidatelist  $\leftarrow$  []
4:   for i in list do
5:     attack, attack_pattern  $\leftarrow$ 
       get_score(chessboard, i, color)
6:     defence, defence_pattern  $\leftarrow$ 
       get_score(chessboard, i, -color)
7:     if attack_pattern and defence_pattern
       can consist of some special pattern then
8:       add extra score for those special chess
       pattern to attack and defence
9:     end if
10:    value  $\leftarrow$  attack + defence
11:    if value > maxscore then
12:      maxscore  $\leftarrow$  value
13:      candidatelist.append(i)
14:    end if
15:  end for
16: end function

```

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### 3. Empirical Verification

For empirical verification, using `code_check_test.py` file to check whether the algorithm works and can give the right answer. The result shows my algorithm works well.

#### 3.1. Design

The main function in this project is `go()` and I implement other four functions to assist `go()` to meet the goal that give a decision where to place the next stone. And among the four functions (except `go` function), `get_stone_around` is an extra function aims at reducing useless computing and decrease the running time.

#### 3.2. Data and data structure

Data used in this project is some `chess_log` files to improve the details and fix bugs. Data structures used include tuple, list, dictionary, array.

#### 3.3. Performance

Instead of using game tree, this AI is simply designed using whole-chessboard-transversal. The time complexity will not be very high, which is about  $\mathcal{O}(n^2)$ . This AI can make decision very quickly compared to game tree algorithm.

#### 3.4. Result

See the score and rank on website **10.20.96.148**. The score suggests my algorithm works not bad.

#### 3.5. Analysis

This AI project is just a simple realization and not even use some advanced algorithms like game tree, Monte Carlo tree. Compared with those AI, this only consider a few situation. Thus, it may perform not so well as those using game tree and it can not handle some special opening like D4, I7. But it is also unreal to implement an AI that can deal with all possible situation by using the tools and data in our hands. In a word, this kind of simple AI's performance absolutely depends on how well you know about Gomoku, and it reminds me that it is a long way to program a true AI.

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### References

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