Programming Pentago with Matlab

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The Objective

- ► To implement an artificial intelligence able to play effectively the board game Pentago.
- ► To use Matlab to achieve so
 - ▶ Most are developed in C++... BUT Matlab has a pedagogical value.
 - Previously used at Supélec's EL.
 - Previously used by our advisor to implement an AI

What is Pentago?

- ► Two-player game introduced by Tomas Flodén in 2005
- ► How do we play it?
 - We play it using a 6x6 board divided into four 3x3 sub-boards (or quadrants).
 - Each player assumes a color (a white player, a black player). Taking turns, each one places a marble of his color onto an unoccupied space on the board, and then rotate one of the sub-boards by 90 degrees, either clockwise or anti-clockwise.
 - A player wins by getting five of their marbles in a vertical, horizontal or diagonal row.
 - If all 36 spaces on the board are occupied without a row of five being formed then the game is a draw.

Why Pentago?

- ▶ Two player, deterministic, perfect knowledge, zero sum game.
 - Interesting for computer science:
 - ▶ Al development and theoretical complexity analysis.
 - Great popularity worldwide:
 - Pentago: smaller number of state and some symmetries.
 - Less complex.
- Pentago has been strongly solved with a Cray supercomputer at NERSC.
 - ▶ How far can we go using Matlab and a regular laptop?
 - Would it be suitable for Human x Machine and Machine x Machine battles?

Work Planning

- First, Pentago structure (i.e. player vs player enabled).
 - Graphical interface.
 - Allow piece placing in an empty space and turn structure.
 - Introduce rotation.
 - Introduction of a «game over» clause (win or tie).
- Second, artificial intelligence (i.e. player vs AI enabled)
 - Random play AI.
 - Research about possible solving algorithms.
- Finally, AI optimizing (i.e. AI vs AI enabled)
 - Implementation of efficient solving algorithms.
 - Further tuning for optimal results.

Graphical components

- Graphical components provided by MATLAB:
 - ▶ figure(...): window.
 - uicontrol(...): two buttons, Start Game et Reset Game, and textarea.
 - ► axes(...): area .
 - rectangle(...): each one of the 36 circles.
 - ▶ image(...): 8 arrows.
 - ▶ line(...): two dotted red lines.

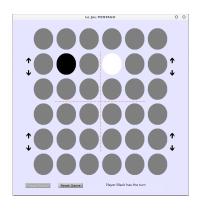


Figure 1: The game's interface.

Positioning of the components

- Parameters of function figure:
 - ▶ **Visible**: determines whether the window will be visible or not.
 - Position: an array containing the coordinates x and y of the screen, width and height of the window, respectively.
 - ▶ Name: the string that will be displayed in the figure window.
 - Number Title: this property set to off means that the string Figure No will not be displayed in the figure window.
 - ▶ Resize: set to off means that the window cannot be resized.
 - Color: sets the background color of the figure. The array contains three values, which determine the color according to the (Red, Green, Blue) scale.
- Example:

Positioning of the components

- Parameters of function uicontrol:
 - Callback: It specifies a function handler, which will be activated when the button is pressed.

- ▶ Parameters of the function **image**:
 - ButtonDownFcn: equivalent to the property Callback detailed previously.
 - ► Function **imread(...)**: reads an image from specified file.

```
rotate_up = imread('arrow_alt_up.jpeg');
textQ1R = image(-30,400,rotate_up, 'ButtonDownFcn', {@turn_pressed, 1, 'L'});
```

Positioning of the components

- Parameters of the function axes:
 - Color: specify the color of the axes back planes. Setting it to none means that the axes is transparent and the figure color shows through
 - Visible: component's visibility. Setting it to off prevents axis lines, tick marks, and labels from being displayed.

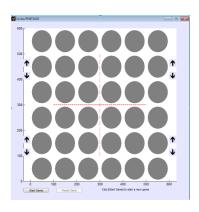


Figure 2: Axis1 with Color and Visible different from none and off

Evaluating a board

- Les variables utilisées:
 - value: the result that will be given to that board.
 - ► Each element of hasBeenDetected is an integer whose representation in binary determines if that element has already been detected taking place in at least one line/column or diagonal.
 - playerScores: an array that stores the scores of all combination the player's got so far in each one of the possible configurations.
 - ▶ **opponentScores**: is the same that *playerScores* but for the opponent.
 - foundUltraCondition is true if we detect a line, column or diagonal completely filled with 5 pieces.

Evaluating a board

► Example, checking the lines for a row:

```
1
    while (offset_index >= 1) && (bitand(hasBeenDetected(x,y), 2) == 0) && ...
2
             (foundUltraCondition == 0)
3
4
         if (y + offset_index) <= 6
             t = state_matrix (x , y:(y + offset_index));
6
7
             if all(t == t(1))
8
a
                 if offset index == 4
10
                     foundUltraCondition = 1:
11
                 end
12
13
                 scores(2, index player) = scores(2, index player) + 10^(3*(offset index+1));
14
                 has Been Detected (x, v:(v + offset index)) = ...
15
16
                     hasBeenDetected (x , y:(y + offset_index)) + 2;
17
18
             elseif offset index == 4
19
20
                 if (sum (t == player) == 4) && (sum (t == opponent) == 1)
21
                     value = value + 10^14:
                 elseif (sum (t == player) == 1) && (sum (t == opponent) == 4)
22
23
                     value = value - 10^14:
24
                 end
25
26
             end
27
         end
28
         offset_index = offset_index - 1;
29
    end
```

The choice of the algorithm

- Computational resources constraints: processing time and memory.
 - Search algorithms with acceptable complexity instead of full search.
 - Ex.: Alpha-beta, Proof-number, Threat-space, Retrograde.
 - Most employed and our choice: Alpha-beta search
- The Alpha-beta search:
 - Based on the minimax algorithm.
 - Two players evaluate a function, one of them tries to maximize and to the other to minimize it.
 - ► The value that will be achieve is: max{min{max{min...max{min{eval}}}}}}
 - Pruning:
 - If a branch cannot change the result, we skip its whole subtree.

Algorithm and complexity

- Algorithm description:
 - Consider a node n somewhere in the tree, such that one can move to that node.
 - 2. If there is a better choice m either at the parent of the node n or at any choice point further up, n will never be reached.
 - Once we have enough information about n to reach this conclusion, we can prune it.
- Complexity, by the arrange of the nodes (branching factor b, search depth d):
 - ▶ Pessimal: $\mathcal{O}(bd)$ the same as the simple minimax.
 - ▶ Optimal: $\mathcal{O}(\frac{bd}{2}) = \mathcal{O}(\sqrt{bd})$.
 - Random: $\mathcal{O}(b\frac{3d}{4})$.

The pruning... Alpha? Beta?

- ▶ Alpha-Beta pruning gets its name from the following parameters:
 - 1. $\alpha =$ the value of the best choice we have found so far at any choice point along the path for MAX.
 - β = the value of the best choice we have found so far at any choice point along the path for MIN.
- Alpha-Beta search:
 - ▶ Updates the values of α and β as it goes along.
 - Prunes the remaining branches at a node
 - As soon as the value of the current node is known to be worse than the current values of α and β.
 - ▶ Therefore, it does not affect the solution.

Demonstration - Video

 Minimax - Alpha Beta Pruning (Artificial Intelligence) by Ice Blended.

Available at: https://youtu.be/SROIGH1P2No?t=104. Start watching at 1:45

Coding (1)

- ▶ The function will return an array. This array will contain:
 - eval_value: the evaluation of the board.
 - move: the place (convention [line column]) where the next piece will be played
 - quadrant: The quadrant where the rotation is applied after the piece be placed.
 - direction: The direction of the rotation.
- Example:

```
function [eval_value move quadrant direction] =
    alpha_beta_search (state_matrix, emptySlots, depth,
    alpha, beta, isMaximizing, player)
```

Coding (2) - Variables and Recursion

- ▶ Variables used in the search:
 - ▶ alpha, beta: Parameters of the Alpha-Beta search, said previously.
 - depth: The depth left by the search. Starts at 2 goes to 0.
 - mustStop: allows us to do the pruning. Do so by breaking the while loops that are used to explore our tree, thus skipping the subtree.
- Recursive function:
 - Base case: maximum depth reached or no possible plays.

Else: explore subtree.

```
else
dir = ['L' 'R']; mustStop = 0;
if (isMaximizing == 1) eval_ = -inf;
else eval_ = +inf;
end
```

Coding (3) - A cascade of loops

- ▶ A cascade of while loops to emule a tree structure. The exploration stops when we arrive to:
 - ▶ Natural limits of the 6x6 board.
 - Pruned subtrees.
- ▶ In our case:

```
1 = 1;
    while (1 <= 6) && (mustStop == 0)
3
         c = 1;
        while (c <= 6) && (mustStop == 0)
             if state_matrix (1,c) == 0
                 quad_index = 1;
7
                 while (quad_index <=4) && (mustStop == 0)
8
                     dir_index = 1;
                     while (dir_index <= 2) && (mustStop == 0)
                          if isMaximizing == 1
10
                              state_matrix (1,c) = player;
11
12
                          else
                              if player == 'B'
13
14
                                  state_matrix (1,c) = 'W';
15
                              else
                                  state_matrix (1,c) = 'B';
16
17
                              end
18
                          end
                          state matrix = rotate quadrant (state matrix, quad index,
19
                               dir (dir index) ):
20
```

Coding (4) - Alpha-beta search coded

- ▶ Implementation of the alpha-beta search algorithm.
 - Started as a pseudo code.

```
state_game_ = verify_victory(state_matrix);
if isMaximizing == 1
    if state game == player
        eval MIN = +10^20:
        mustStop = 1:
    else
        [eval MIN move MIN quadrant MIN
              direction MIN1 = ...
            alpha beta search(state matrix.
                   emptvSlots - 1, depth - 1,
                  alpha, beta, 0, player);
    end
    if eval < eval MIN
        eval_ = eval_MIN;
        eval value = eval :
        move = [1 c]:
        quadrant = quad index:
        direction = dir (dir index):
    end
    if eval >= beta
        mustStop = 1;
    end
    alpha = max(alpha, eval_)
```

```
elseif isMaximizing == 0
   if (state_game_ ~= player) && (state_game_
          ~= 0) && (state game ~= 'D')
       eval MAX = -10^20:
       mustStop = 1:
        [eval MAX move MAX quadrant MAX
              direction MAX1 = ...
            alpha beta search(state matrix.
                  emptvSlots - 1, depth - 1,
                  alpha, beta, 1, player);
   end
    if eval > eval MAX
       eval = eval MAX:
       eval value = eval :
       move = [1 c]:
       quadrant = quad index:
       direction = dir (dir index):
        end
    if alpha >= eval
       mustStop = 1:
    beta = min(beta, eval_);
```

Coding (5) - End of an iteration

- Restore the state of our board(state_matrix) to how it was before doing the play that we are analysing
 - Achieved by doing the reverse of all modifications that we have applied since we ran the alpha beta search on the board.
- ▶ Update loop parameters and start another iteration.

```
if dir (dir_index) == 'L'
1
                    state_matrix = rotate_quadrant(
                        state_matrix, quad_index, 'R');
           else
3
                    state_matrix = rotate_quadrant(
                        state_matrix, quad_index, 'L');
           end
           state_matrix (1,c) = 0;
7
           dir index = dir index + 1:
   end
   quad_index = quad_index + 1;
10
   end
11
```

Simulation

- Player Black represents your artificial intelligence.
- Player White represents ours.
- ▶ 500 matches simulated:
 - Each AI has the first move in exactly half of them, that is, 250 matches.
 - 193 (90 of them when it started playing) were won by player White.
 - 223 (116 of them when it started playing) were won by player Black.
 - ▶ 84 (44 when **Black** started) resulted in draws.

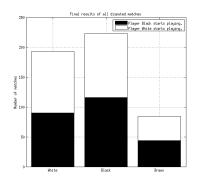


Figure 3: Number of victories of each player.

Some probabilities

- Based on the previous results, we can observe that:
 - For player Black:

$$P(\mathbf{B}) = \frac{223}{500} = 0.446$$

 $P(\mathbf{B} \mid \mathbf{B}_{started}) = \frac{116}{250} = 0.464$
 $P(\mathbf{B} \mid \mathbf{W}_{started}) = \frac{107}{250} = 0.428$

For player White:

$$P(W) = \frac{193}{500} = 0.386$$

 $P(W \mid B_{started}) = \frac{90}{250} = 0.360$
 $P(W \mid W_{started}) = \frac{103}{250} = 0.412$

▶ In case of match nul:

$$P(M_{draw}) = \frac{84}{500} = 0.168$$

 $P(M_{draw} \mid B_{started}) = \frac{44}{250} = 0.176$
 $P(M_{draw} \mid W_{started}) = \frac{40}{250} = 0.160$

Decision time for each player

▶ Player Black takes, in average,2.05 seconds to decide.

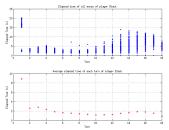


Figure 4: Required time to Player **Black** take a move.

Player White takes, in average,9.5 seconds to decide.

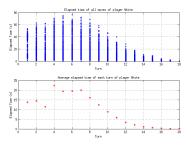


Figure 5: Required time to Player White take a move.

Comparison of algorithms

- ▶ The difference between these two times is explained by how which algorithm decides the move and evaluates a board.
 - Our algorithm needs to verify each position of the matrix at least twice: once to evaluate if that position belongs to a row, column or diagonal, and another to conclude if that move is profitable.
- The first eight moves of player White take almost all the time of decision.
 - That's already expected: at the beginning of the match, there are more positions to be verified, contrarily to the end, in which few available positions remain.
- ▶ Player **Black**'s moves takes roughly constant time, around 2 seconds, to be decided. The only move that does not follow that property is the first one.

Conclusion

- Obtained results:
 - Graphical interface: Easy to use, functional and enjoyable.
 - AI:
 - Human vs Human. Al vs Al and mixed matches enabled.
 - Good performance: Comparable with the one of our advisor.
- ▶ Room for improvement (some of them may be beyond a 2 month project!):
 - Programming Language: C++ instead of Matlab.
 - Algorithm: Transition tables, deeper search.
 - Hardware: GPU processing, supercomputers.
- Aftermatch
 - The next generation of students may take from where we left and pursuit one of the lines of improvement.
 - Or simply have our algorithm as a reference, as we had Mr. Bourgois'.



Bibliography (1)

- Pentago, (2015). Pentago. [online] Available at: http://en.wikipedia.org/wiki/Pentago [Accessed 14 Apr. 2015].
- Pentago is a first player win, (2015). [online] Available at: https://perfect-pentago.net/details.html#intro [Accessed 14 Apr. 2015].
- ➤ On solving Pentago, (2015). [online] Available at: http://www.ke.tu-darmstadt.de/lehre/arbeiten/bachelor/ 2011/Buescher_Niklas.pdf [Accessed 14 Apr. 2015].
- Standford University, (2015). [online] Available at: http://web.stanford.edu/~msirota/soco/alphabeta.html [Accessed 14 Apr. 2015].
- ▶ Élagage alpha-beta, (2015). [online] Available at: http://fr.wikipedia.org/wiki/%C3%89lagage_alpha-beta [Accessed 14 Apr. 2015].

Bibliography (2)

- ► Chalmers GÖTEBORGS UNIVERSITET, (2015). [online] Available at: http://www.cse.chalmers.se/edu/year/2014/course/TIN172/example-exam-solutions.html [Accessed 14 Apr. 2015].
- McCarthy, 2006. McCarthy, John (LaTeX2HTML 27 November 2006). "Human Level AI Is Harder Than It Seemed in 1955". http://www-formal.stanford.edu/jmc/slides/wrong/wrong-sli/wrong-sli.html [Acessed 20 Dec. 2006]
- Russell, Stuart J., 2003. Russell, Stuart J.; Norvig, Peter (2003), Artificial Intelligence: A Modern Approach (2nd ed.), Upper Saddle River, New Jersey: Prentice Hall, ISBN 0-13-790395-2 http://aima.cs.berkeley.edu/
- ▶ Nersc, (2015). [online] Available at: https: //www.nersc.gov/users/computational-systems/edison [Accessed 19 May 2015].
- ► NVIDIA, (2015). [online] Available at: https://developer.nvidia.com/how-to-cuda-c-cpp [Accessed 19 May 2015].