# **Semester Project**

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#### **Abstract**

In the project, we built the game of 2048 from scratch, developed the logic of the game, built the user interface, and developed an AI that would try to solve this game.

The project is written in Python.

#### **Keywords**

2048 — expectimax — Tkinter

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# **Contents General Guidelines** 1 **Rules and Constraints Heuristics and Algorithm** Dealing with random pop-ups • One-player game • Logics behind 3.3 Running time and consumed space ..........2 2 Challenges Expectimax • Random assignments of numbers • Coordinating all the files **Experiments and Results Summary and Conclusions** 6.1 Summary ......3 3 **Acknowledgments**

# 1. General Guidelines

The project contains three parts: the code, the report and the poster. Below is the detailed description of how this project is constructed and how to run this game.

# 1.1 Code

# • 2048.py

This file contains the GUI to interact with a player. It uses Tkinter to draw labels and boards on window and provides a way to interact with the player, either by clicking on the 'New Game' button or the 'AI Player' button.

Save 2048.py, Board.py and AIPlayer.py under the same directory. Run the module in 2048.py, and a window would pop up. There are three frames in this file: the top frame, the board frame, and the bottom frame.

Top Frame: The icon of the game, '2048' would show in the upper left corner of the window, with the current score and the best score showing to its right. This frame also contains the buttons for a new game or for the AI player.

Board Frame: This frame contains the actual board of the game.

Bottom Frame: This frame displayers instructions on how to play this game.

#### • Board.py

This file contains the fundamentals of this game. With methods to make moves, check the state of the game, etc, Board.py builds the foundation of the game.

#### AIPlayer.py

This file contains the class for the AI player. The project uses a heuristic to calculate the best score of a board, and by using expectimax algorithm, the project is able to make moves to get the best result.

Detailed explanation would be included in this paper below.

#### 1.2 Others

#### Poster

This poster is used for presentation and demonstration.

#### • Report

This report is to explain in details how this project was initiated and constructed, including instructions on how to run the code.

### 2. Rules and Constraints

#### 2.1 Rules

Please refer to the previous part (General Guidelines) for rules of this game.

#### 2.2 Constraints

The main constraint of building an AI for 2048 is to deal with randomly generated 2s and 4s, on random positions on board.

Another obstacle is to come up with a heuristic that would guarantee to solve this game. As many may already know, the game has a rule to follow and many approaches have been suggested. To find the most efficient heuristic is very important in solving this game.

# 3. Heuristics and Algorithm

# 3.1 Heuristic

# 3.2 Algorithm - Expectimax

#### 3.2.1 Dealing with random pop-ups

Instead of using Alpha-Beta pruning and Minimax algorithm, we chose Expectimax was a way to solve this problem.

Expectiminimax deals with problems with random difficulties. In this project, the player cannot decide if a 2 or 4 would be inserted to the board, and also no way to know where on the board it would be put. Expectiminimax thus has become our first choice to solve this problem.

#### 3.2.2 One-player game

As we were solving Connect-4 game, there are two players in this game. In this project, only one player plays 2048. As min and max scores are used to compare situations suitable for each player, we only need to use the max score in order to make the best move for our player.

#### 3.2.3 Logics behind it

We break the problem into two parts: when it's time for the player to make the move, and when it's time for the board to generate a 2 or a 4 in a random position.

# Player's time

As there are four directions, we make four children boards and calculate the scores for each of the child. We keep record of the best score so far, and only update it when a higher score comes up.

We also keep track of the move in order to get this score.

#### Board generating 2/4

When it is time for the board to put a 2 or a 4, we simulate all the situations with all the possibilities.

For each empty cell on board, we would generate a 2 to put it there, and calculate the score by multiplying the one we get from heuristics and its weight, which is 0.9.

We then repeat this step, but instead of putting 2 on board, we put a 4 there, and the weight now is 0.1. After obtaining both scores for the two conditions, we add them up and would keep a record of the score. In the end, we get the weighted average by dividing the sum by the number of empty cells.

## 3.3 Running time and consumed space

# 4. Challenges

# 4.1 Building the Game

Even though the game is fairly easy and simple, it still took us some time to build this game from scratch. Instead of borrowing from existing code, we first constructed the board class (Board.py). The reason is that this way, we would be able to work on the AI part.

The little challenge we encountered during this process was how to merge the tiles to one direction. There is hardly any better way than to break the matrix into rows and merge the tile row by row, during which we need to sometimes transpose the matrix in order to get all four directions.

As for the GUI, since none of us had experiences in building GUI with Python, we had to learn from the beginning, from the layout to the mouse events and keyboard events.

One challenge in particular that has not been solved is to show how the AI gets to the final board step by step. Though the game itself works perfectly fine, we had little time left to improve our GUI with our AI.

#### 4.2 Implementing the AI

#### 4.2.1 Expectimax

The AI is the hard part of this project, and also the core of our purpose. Understanding expectimax was not hard, since it is very similar to minimax which we already implemented. However, it was a big challenge for us to figure out a way to consider all the random boards generated.

## 4.2.2 Random assignments of numbers

In the end, we realised that the only way to find the best move was to take all the possibilities into account. We used for loops to consider these: all the random positions and both of the situations in which a 2 or a 4 is put on board.

#### 4.2.3 Coordinating all the files

Moreover, coordinating three classes could take a very long time. Though this was not a technical problem, it was very time-consuming to debug the code.

# 5. Experiments and Results

- 5.1 Experiments
- 5.2 Results
- 5.3 Comparison to others

# 6. Summary and Conclusions

- 6.1 Summary
- 6.2 Alternative Approaches

# **Acknowledgments**

This is the final project for CSCI-B 351, Introduction to Artificial Intelligence. The professor is **Professor Saul Blanco**.

We would like to thank the professor and all the teaching assistants for their help throughout the semester. They have provided tremendous help to us and have been greatly helpful. We would like to especially thank our two mentors, **Daniel Iglarsh** and **Ben Lewis**, who kindly gave us time and advice during this project.

(Other teaching assistants are **Jack Langston**, **Tristan Rogers**, **Scott Mathews**, **Kyle Yohler**, **Jacob Lin**, **and Matthew Ploetz**. All of the instructores dedicated their time to help us in office hours and we truly appreciate it.)

Moreover, we would like to thank **Gabriele Cirulli** who created this game. Even though we used a very different way to implement this game, all the credits of this game go to him.