Project Report

Project Title:
Pac Ludo Fusion
Submitted By:
Kainat Faisal (22k-4405) Laiba Khan (22k-4610)
Course:
Artificial Intelligence (AI)
Instructor:
Miss Ravia Ejaz Miss Alina Arshad
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1. Executive Summary

Project Overview:

Pac Ludo Fusion is an innovative reimagination of the classic Pac-Man game. This project integrates Al-driven enemies with a rotating hexagonal maze to enhance gameplay complexity. The main goal was to implement three Al strategies—Minimax, A* Pathfinding, and Q-Learning—to simulate intelligent ghost behavior. Players must collect pellets and survive as the maze dynamically shifts every 15 seconds, requiring constant adaptation and strategic movement.

2. Introduction

Background:

Pac-Man is a classic arcade game centered around navigating a maze, collecting pellets, and avoiding ghosts. This project introduces several enhancements to the original mechanics: a hexagonal grid for greater path diversity and dynamic maze rotation for added challenge. The goal was to explore AI behaviors in a non-trivial, constantly evolving environment.

Objectives of the Project:

- Develop a playable maze game with Al-driven enemy behaviors.
- Implement three distinct ghost Als using Minimax, A* Pathfinding, and Q-Learning.
- Evaluate and compare the effectiveness of these AI techniques.
- Introduce rotating maze mechanics to encourage player adaptability and strategic planning.

3. Game Description

Original Game Rules:

In traditional Pac-Man, the player navigates a fixed maze, eats pellets to score points, and avoids ghosts. Consuming special pellets can temporarily turn the tables, allowing the player to defeat ghosts.

Innovations and Modifications:

- Hexagonal Maze Layout Replaces the conventional square grid with a hex grid for more complex navigation.
- Maze Rotation Rotates every 15 seconds, changing viable paths and strategies.
- Smart Ghosts -

- o **Minimax Ghost**: Uses predictive modeling to intercept the player.
- A Ghost*: Calculates the shortest path to the player.
- Q-Learning Ghost: Learns from experience using reinforcement learning.

Power-Ups –

- Temporary invincibility upon spawn and life loss.
- Bonus life after collecting every 50 pellets.

4. Al Approach and Methodology

Al Techniques Used:

- **Minimax Algorithm** Simulates player and ghost moves to anticipate the most advantageous path.
- A Pathfinding* Computes shortest paths using a heuristic suited to hex grids.
- **Q-Learning (Reinforcement Learning)** Employs a neural network (via TensorFlow) that learns optimal behavior over time.

Algorithm and Heuristic Design:

- Minimax uses depth-limited search with state evaluation based on proximity and threat.
- **A*** leverages a hex-based distance function for accuracy in dynamic pathfinding.
- **Q-Learning** utilizes a reward matrix and state-action table, training through gameplay feedback.

Al Performance Evaluation:

- Minimax Ghost showed strong anticipatory behavior, effectively predicting player moves.
- A Ghost* offered fast, efficient pursuit with low computation time.
- Q-Learning Ghost demonstrated significant improvement with training, adapting well to changes over time.

Each ghost's behavior was evaluated based on capture success rate, path efficiency, and response time to the rotating maze.

5. Game Mechanics and Rules

Modified Game Rules:

- Players begin with 5 lives.
- The maze rotates every 15 seconds.
- Players collect **pellets for points** and can earn extra lives.

Turn-Based Mechanics:

- Gameplay is real-time, controlled using **W**, **A**, **S**, **D** keys.
- Players can restart the game after victory or defeat by pressing **R**.

Winning Conditions:

- Victory: Collect all pellets.
- Game Over: Lose all 5 lives.

6. Implementation and Development

Development Process:

The game was developed in Python using the **Pygame** library. Core gameplay, UI, and AI logic were structured into object-oriented modules. Each ghost type was implemented in a modular AI engine that interacts seamlessly with the game environment.

Programming Languages and Tools:

• Language: Python

• Libraries: Pygame, NumPy, TensorFlow

• **Tools:** GitHub (version control), PyCharm (IDE)

Challenges Encountered:

• Implementing rotation for a hex grid without disrupting pathfinding logic.

• Optimizing performance with three concurrent AI systems.

• Training the Q-Learning agent effectively in a real-time, dynamic maze.

7. Team Contributions

Kainat Faisal

- Designed the hexagonal maze system.
- Implemented the A* pathfinding ghost.
- Co-developed the Q-learning agent and game mechanics.

Laiba Khan

- Designed and implemented the Minimax-based ghost logic.
- Worked on decision heuristics and AI evaluations.
- Co-developed the Q-learning agent and game mechanics.

8. Results and Discussion

Al Performance Overview:

- Minimax Ghost:
 - Success Rate: 85%
 - Strong predictive capabilities.
- A Ghost*:
 - Decision Time: ~10ms
 - Most efficient pathing algorithm.
- Q-Learning Ghost:
 - Initial Success Rate: 30%
 - Post-training Success Rate: 65%
 - o Performance improves significantly with training and adaptation.

The rotating maze added dynamic complexity, requiring the AI systems to be robust and flexible. The A* algorithm remained consistently reliable, while Q-Learning adapted best over time.

9. References

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