Project Title: Quantum Snakes & Ladders - AI Enhanced Board Game

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1. Project Overview

Project Topic:

Quantum Snakes & Ladders is a modified version of the classic Snakes & Ladders board game. The game incorporates **strategic AI**, **power-ups**, **and quantum elements** to increase complexity and engagement. The game will feature:

- Al-controlled players using Minimax with Alpha-Beta Pruning for decision-making.
- Power-ups like Reroll, Superposition, and Entanglement to influence gameplay.
- Monte Carlo Tree Search (MCTS) for handling uncertainty in dice rolls.

Objective:

- Develop an AI strategy that optimizes dice roll choices using Minimax with Alpha-Beta Pruning.
- Implement a new set of rules and power-ups to enhance game complexity.
- Use **MCTS** to simulate long-term game outcomes for decision-making.
- Develop a graphical interface (GUI) using Pygame for better gameplay experience.

2. Game Description

Original Game Background:

Snakes & Ladders is a **classic board game** where players roll a die to move forward on a numbered board. Landing on a ladder moves a player up, while landing on a snake moves them down. The goal is to reach the final square first.

Innovations Introduced:

- **AI-powered players:** The game includes an **AI opponent** that makes intelligent decisions instead of relying solely on random dice rolls.
- **Power-ups:** Players can collect power-ups that allow them to:
 - Reroll: Get another chance to roll the die.
 - > **Superposition**: Choose between two possible dice rolls.
 - > **Entanglement**: Swap positions with another player.
- **Strategic AI-based dice selection:** Instead of random dice rolls, the AI chooses the best roll using Minimax.
- Monte Carlo simulations: The AI predicts the best long-term moves using MCTS.

3. Al Approach and Methodology

Al Techniques to be Used:

- Minimax Algorithm: Determines the best move by assuming all players play optimally.
- Alpha-Beta Pruning: Optimizes Minimax to reduce unnecessary computations.
- Monte Carlo Tree Search (MCTS): Simulates multiple dice rolls and predicts the best move in uncertain scenarios.
- Reinforcement Learning (Optional): Al learns optimal strategies through self-play.

Heuristic Design:

- Avoiding snakes and aiming for ladders for positional advantage.
- **Evaluating power-ups** and deciding the best time to use them.
- **Predicting opponent moves** to minimize their advantage.

Complexity Analysis:

- Minimax Time Complexity: O(b^d) (where b = branching factor, d = depth of decision tree)
- Alpha-Beta Pruning: Reduces complexity to O(b^(d/2)).

 Monte Carlo Tree Search: Balances computation with randomized simulations for practical efficiency.

4. Game Rules and Mechanics

Modified Rules:

- Players can **collect power-ups** instead of rolling the die.
- Al-controlled players use intelligent dice selection instead of random rolls.
- Certain board positions trigger **special events** like teleportation or power-up loss.

Winning Conditions:

- A player wins by reaching the final square first.
- If multiple players reach the final square in the same turn, the player with **fewer total** moves wins.

Turn Sequence:

- Each player rolls a six-sided die (1-6) (Al chooses the best roll instead of random rolling).
- The player moves accordingly and activates any board effects (snakes, ladders, power-ups).
- If a power-up is available, the player can use or save it for later.
- The turn passes to the next player.

5. Implementation Plan

Programming Language:

Python

Libraries and Tools:

- **Pygame:** For board visualization and user interaction.
- **NumPy:** For handling data and AI calculations.
- **Scikit-learn:** For implementing reinforcement learning (if applied).
- TensorFlow (Optional): For training an advanced AI model.

Milestones and Timeline:

Week Task

- 1-2 Game design, board layout, and rule finalization
- 3-4 Al strategy development (Minimax, Alpha-Beta Pruning)
- 5-6 Coding core game mechanics and player movement
- 7 Al integration and testing with MCTS
- 8 Final testing, debugging, and report preparation

6. References

- "Artificial Intelligence: A Modern Approach" by Stuart Russell & Peter Norvig.
- Research papers on Minimax, MCTS, and AI for board games.
- Pygame documentation: https://www.pygame.org/docs/
- Monte Carlo Tree Search in Board Games: A Survey.