Project Report

Project Title: AI-Powered Strategic Checkers Game Using Minimax Algorithm

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Course: AI

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1. Executive Summary

• Project Overview:

This project focuses on developing an AI-powered version of the traditional Checkers game. The core objective is to build a strategic AI opponent using the Minimax Algorithm to simulate intelligent gameplay. The AI analyzes current board states, predicts future outcomes, and makes optimal moves based on a heuristic evaluation function. Innovations include board state heuristics, AI decision-making, and optional Alpha-Beta pruning to optimize performance.

2. Introduction

• Background:

Checkers is a classic two-player game played on an 8x8 grid, however, here it is done on a 10x10 grid, where players move diagonally and aim to capture all opponent pieces. This project was chosen to explore strategic decision-making in a deterministic game. To enhance gameplay, an AI component has been integrated using the Minimax Algorithm, enabling dynamic, intelligent opposition for solo players.

• Objectives of the Project:

- 1. Implement a playable version of Checkers with a responsive UI.
- 2. Develop an AI using the Minimax Algorithm.
- 3. Introduce a heuristic evaluation function for game states.
- 4. Optimise performance with optional Alpha-Beta pruning.
- 5. Test the AI against human players for effectiveness and difficulty.

3. Game Description

• Original Game Rules:

1. Played on an 8x8 board with alternating dark and light squares.

	Each player begins with 12 pieces on the dark squares of the first three rows. Pieces move diagonally forward and capture by jumping. Kings (promoted pieces) can move backward as well. The game ends when a player loses all pieces or can no longer move. Innovations and Modifications: Al opponent using Minimax to determine the best move. Board evaluation considers piece count, kings, and positioning. Move simulation enables lookahead strategy. Forced capture rule enforced programmatically.
	Optional Alpha-Beta pruning for performance enhancement.
4. AI A	Approach and Methodology
• 1	AI Techniques Used:
	Minimax Algorithm: Evaluates all possible game states recursively to choose the optimal move.
	☐ Alpha-Beta Pruning (Optional): Reduces the number of nodes evaluated in Minimax by pruning unneeded branches.
• 1	Algorithm and Heuristic Design:
	1. Piece Value: Regular = 1 point, King = 3 points.
	2. Positional Advantage: Points for center control and safe positions.
	3. Capture Opportunities: Moves leading to captures are prioritized.4. The evaluation function combines these metrics to assign a score to each board state.
• 1	AI Performance Evaluation:
	\square AI tested against human players of varying skill.
	\square Metrics: win rate, average decision time, and number of moves considered.
5. Gar	ne Mechanics and Rules
•]	Modified Game Rules:
	\square Single-player mode where AI plays as one of the players.
	\square Forced capture if a jump is available.
	☐ Promotion to King when a piece reaches the opponent's baseline.

\square Kings can move both forward and backward.
\square On a 10x10 grid
• Turn-based Mechanics:
\square Turns alternate between player and AI.
\square After each player move, the AI evaluates and makes its move
automatically
• Winning Conditions:
A player wins if the opponent has no legal moves or all pieces are captured.
6. Implementation and Development
• Development Process:
☐ Designed game logic and UI using Python.
\square Implemented Minimax algorithm with recursive move evaluation.
☐ Integrated board evaluation heuristics.
☐ Conducted iterative testing and debugging.
• Programming Languages and Tools:
☐ Programming Language: Python
Libraries: Pygame (for game interface), NumPy (for matrix handling, optional)
☐ Tools: GitHub (version control)
• Challenges Encountered:
\square Ensuring efficient move generation and backtracking for Minimax.
Preventing performance lags during deeper searches.
☐ Balancing AI strength and response time.
7 Toom Contributions
7. Team Contributions
 Team Members and Responsibilities: [Taimoor]: Implemented Minimax and Alpha-Beta pruning.
[Rafey]: Developed game UI using Pygame.
[Urwa]: Worked on heuristic function design and board evaluation logic.

8. Results and Discussion

• AI Performance:

The AI demonstrated consistent strategic play and outperformed beginner
human players in over 80% of test games.
Decision time averaged 1.5–3 seconds per move with Alpha-Beta pruning.
The heuristic evaluation effectively guided decision-making, especially in
endgame scenarios.

9. References

- Russell, S., & Norvig, P. (2020). Artificial Intelligence: A Modern Approach.
- Checkers game rules: Wikipedia
- Minimax and Alpha-Beta pruning tutorials: GeeksforGeeks, Stack Overflow