
AlphaDeepChess: chess engine based on alpha-beta pruning



Final Degree Project
Course 2024–2025

Authors

Juan Girón Herranz

Yi Wang Qiu

Directors

Ignacio Fábregas Álfaro

Rubén Rafael Rubio Cuéllar

Bachelor's Degree in Computer Engineering

Faculty of Computer Science

Complutense University of Madrid

Bachelor's Degree in Videogame Development

Faculty of Computer Science

Complutense University of Madrid

AlphaDeepChess: chess engine based on alpha-beta pruning

Final Degree Project in Computer Engineering
Final Degree Project in Videogame Development

Authors

Juan Girón Herranz
Yi Wang Qiu

Directors

Ignacio Fábregas Álfaro
Rubén Rafael Rubio Cuéllar

Convocation: *June 2025*

Bachelor's Degree in Computer Engineering
Faculty of Computer Science
Complutense University of Madrid

Bachelor's Degree in Videogame Development
Faculty of Computer Science
Complutense University of Madrid

March 26, 2025

Dedictory

*To our younger selves, for knowing the art of
chess*

Acknowledgments

To our family members for their support and for taking us to chess tournaments to compete.

Abstract

AlphaDeepChess: chess engine based on alpha-beta pruning

Chess engines have significantly influenced the development of computational strategies and game-playing algorithms since the mid-20th century. Computer scientists as renowned as Alan Turing and Claude Shannon set the foundations for the development of the field. Thereafter, hardware and software improvements and the evolution of heuristics would build upon these foundations, including the introduction of alpha-beta pruning, an optimization of the minimax algorithm that significantly reduced the number of nodes evaluated in a game tree. With increasing computational power, modern engines such as Stockfish or Komodo leverage not only search optimizations but also advancements in heuristics and, in some cases, artificial intelligence using neuronal networks.

Keywords

chess engine,alpha-beta pruning,minimax algorithm,game tree search,heuristic evaluation,move ordering,search optimization,transposition tables,zobrist hashing

Contents

1	Introduction	1
1.1	Objetives	1
1.2	Work plan	2
2	Status of the art	3
3	Work description	5
3.1	Where to begin?	5
3.2	Basic concepts	5
3.2.1	Chess board	5
3.2.2	Chess pieces	5
3.2.3	Movement of the pieces	5
3.2.4	Rules	5
3.3	Modules	6
3.3.1	Board	6
3.3.2	Move generator	6
3.3.3	Move ordering	6
3.3.4	Evaluation	6
3.3.5	Search	6
3.4	Code implementation	6
3.4.1	Data representation	6
3.4.2	Initialized memory	6
3.5	Other	6
3.5.1	Board visualizer using Python	6
3.5.2	Comparation using Cutechess and Stockfish engine	6

3.5.3 Profiling	6
4 Conclusiones y Trabajo Futuro	7
Personal contributions	9
A Título del Apéndice A	11
B Título del Apéndice B	13

List of Figures

List of Tables

Chapter 1

Introduction

“The most powerful weapon in chess is to have the next move”

— David Bronstein

Chess, one of the oldest and most strategic games in human history, has long been a domain for both intellectual competition and computational research. The pursuit of creating a machine that could compete with the best human players, chess Grandmasters (GM), was present. It was only a matter of time before computation surpassed human computational capabilities.

In 1997, the chess engine Deep Blue made history by defeating the world champion at the time, Garry Kasparov, marking the first time a computer had defeated a reigning world champion in a six-game match under standard chess tournament time controls¹.

Since then, the development of chess engines has advanced rapidly, moving from rule-based systems to AI-driven models. However, classical search algorithms, such as alpha-beta pruning, continue to be fundamental to understanding the basics of efficient search and evaluation of game trees.

1.1 Objectives

- Develop a functional chess engine using alpha-beta pruning as the core search algorithm.
- Optimize search efficiency by implementing move ordering, quiescence search, and iterative deepening to improve pruning effectiveness.
- Implement transposition tables using Zobrist hashing to store and retrieve previously evaluated board positions efficiently.

¹[https://en.wikipedia.org/wiki/Deep_Blue_\(chess_computer\)](https://en.wikipedia.org/wiki/Deep_Blue_(chess_computer))

- Ensure modularity and efficiency so that the engine can be tested, improved, and integrated into chess-playing applications.
- Compare performance metrics against other classical engines to evaluate the impact of implemented optimizations.

1.2 Work plan

1. Research phase and basic implementation: understand the fundamentals of alpha-beta pruning with minimax and position evaluation. Familiarize with the UCI (Universal Chess Interface) and implement the move generator with its specific exceptions and rules.
2. Optimization: improve search efficiency using transposition tables and Zobrist hashing.
3. Comparison: use Stockfish to compare efficiency generating tournaments between chess engines and a profiler to detect possible bottlenecks.

Chapter 2

Status of the art

Chapter 3

Work description

3.1 Where to begin?

Explain how is organised the different modules of the code and what is UCI in order to help the current reader to recreate a mental map of how to use it.

3.2 Basic concepts

Introduction to the basic concepts of chess.

3.2.1 Chess board

3.2.2 Chess pieces

3.2.3 Movement of the pieces

3.2.4 Rules

White starts first, possible results (win, draw or lose), 30 moves rule, zugzwang?, check, checkmate, ...

3.3 Modules

3.3.1 Board

3.3.2 Move generator

3.3.3 Move ordering

3.3.4 Evaluation

3.3.5 Search

3.4 Code implementation

3.4.1 Data representation

Use of `uint64_t` as bitboards to store the board information and other code structures and classes justifying why is efficient.

3.4.2 Initialized memory

Some tables are memory initialized instead of computed, explain it.

3.5 Other

3.5.1 Board visualizer using Python

3.5.2 Comparison using Cutechess and Stockfish engine

3.5.3 Profiling

Chapter 4

Conclusions and Future Work

Conclusiones del trabajo y líneas de trabajo futuro.

Antes de la entrega de actas de cada convocatoria, en el plazo que se indica en el calendario de los trabajos de fin de grado, el estudiante entregará en el Campus Virtual la versión final de la memoria en PDF.

Personal contributions

Student1

Al menos dos páginas con las contribuciones del estudiante 1.

Student 2

Al menos dos páginas con las contribuciones del estudiante 2.

Appendix **A**

Título del Apéndice A

Los apéndices son secciones al final del documento en las que se agrega texto con el objetivo de ampliar los contenidos del documento principal.

Appendix	B
----------	----------

Título del Apéndice B

Se pueden añadir los apéndices que se consideren oportunos.

Este texto se puede encontrar en el fichero Cascaras/fin.tex. Si deseas eliminarlo, basta con comentar la línea correspondiente al final del fichero TFGTeXIS.tex.

*–¿Qué te parece desto, Sancho? – Dijo Don Quijote –
Bien podrán los encantadores quitarme la ventura,
pero el esfuerzo y el ánimo, será imposible.*

*Segunda parte del Ingenioso Caballero
Don Quijote de la Mancha
Miguel de Cervantes*

*–Buena está – dijo Sancho –; fírmela vuestra merced.
–No es menester firmarla – dijo Don Quijote–,
sino solamente poner mi rúbrica.*

*Primera parte del Ingenioso Caballero
Don Quijote de la Mancha
Miguel de Cervantes*

