



# What computers teaches us about chess

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# Overview

- History
- Current chess engines
- Evaluation function
- Building a search tree



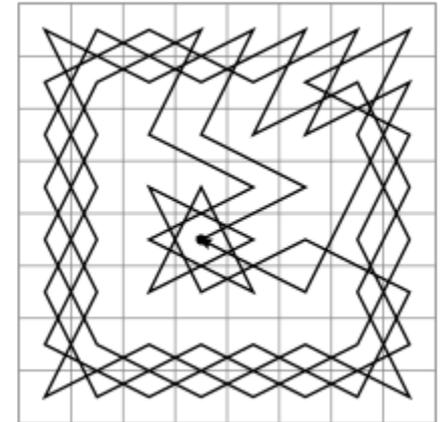
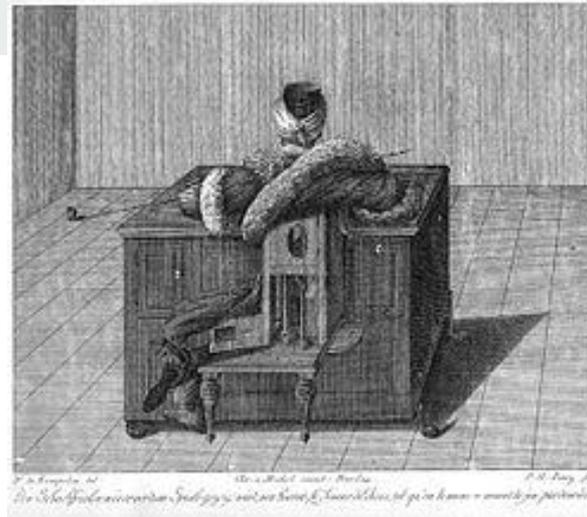
# History



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# 1769

- Wolfgang von Kempelen builds the The Turk (Automaton Chess Player)
- Gave the illusion that it was an intelligent playing machine, but was in fact operated by a skilled operator.
- Won many games in Europa and the Americas, including games against Napoleon Bonaparte and Benjamin Franklin.





## 1950's

- 1950 - Claude Shannon publishes “Programming a Computer for Playing Chess” - one of the first papers on the problem of computer chess.
- 1951 - Alan Turing publish the first program, developed on paper, capable of playing a full game of chess. Known as Turochamp.
- 1956 - Los Alamos chess is the first “chess-like” program, playing on a MANIAC I computer
- 1956 - John McCarthy develops the alpha-beta search algorithm.
- 1957 - First programs to play a complete game of chess developed by Alex Bernstein.



## 1960's

- 1958 - First program to play credibly, Kotok-McCarthy published by MIT
- 1963 - Grandmaster David Bronstein defeats M-20 running an early chess program.
- 1966-67 - First chess match between computers. Moscow Institute for Theoretical and Experimental Physics defeats Kotok-McCarthy at Stanford University by telegraph over nine months.



## 1970's

- 1974 - The first World Computer Chess Championship is won by Russian program Kaissa.
- 1975 - First commercial microcomputer, the Altair 8800.
- 1976 - First dedicated commercial chess computer, Chess Challenger released by Fidelity Electronics.
- 1976 - Microchess is the first commercial software for a microcomputer.



## 1980's

- 1981 - Cray Blitz wins Mississippi State Championship with a perfect score of 5-0. First computer to defeat a master.
- 1981 - IBM releases the PC.
- 1986 - Chessmaster 2000 released - to become the world's best selling line of chess programs.
- 1987 - Chessbase founded by Frederic Friedel and Matthias Wullenweber - the first chess database program.
- 1988 - Deep Thought becomes the first computer to beat a GM in a tournament, performance of 2745
- 1989 - Deep Thought loses two exhibition matches against Gary Kasparov.



# ACM Chess Challenge

Garry Kasparov  
VS



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## 1990's

1992 - ChessMachine wins the World Computer Chess Championship. It is the first time a microcomputer beats a mainframe.

1993 - Deep Blue loses a four game match against Bent Larsen

1996 - Deep Blue loses a six-game match against Garry Kasparov.

1997 - Deep Blue wins a six-game match against Garry Kasparov.





## 2000's

- Early 2000's - Various matches between computers and humans: Kramnik vs Deep Fritz, Kasparov vs Deep Junior, Kasparov vs X3D Fritz, Hydra vs Adams, etc.
- 2005 - Rybka wins IPCCC tournament and quickly becomes the strongest engine.
- 2008 - Stockfish 1.0 released in November 2008

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## 2010's



- 2014-2019. Stockfish won various computer world chess championships.
- 2019. Stockfish 10 is the world's strongest publicly available chess engine - it is free and open source.
- 2017 - AlphaZero from Google beats Stockfish 28-0, with 72 draw in a 100 game match.
- 2017 - AlphaZero taught itself to play chess using 5000 TPUs using neural networks. After 4 hours, it could play stronger than Stockfish 8 and after 9 hours of training, beat Stockfish in a controlled 100-game tournament.
- 2018 - AlphaZero paper published in journal Science on 7 December 2018.



# How strong are chess computers?

- Current estimates for the strength of the best chess engines are at about 3300 to 3500.
- Magnus Carlsen currently has a rating of 2845.
- This roughly means that the best humans could score about 2 to 4 games in a 100 game match.

**“Playing against the computer is like running against a Ferrari”**

**The beginning and the end...**

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# Opening Books

- All chess engines have extensive opening books from which it makes its first moves.
- In each position, there are a number of moves that can be played.
- Each move has an associated probability - the engine makes a move based on these probabilities.
- Once the moves in the opening book is depleted, it starts making moves based on evaluation.

## LESSON:

- You simply **NEED TO KNOW** your opening lines to play them accurately like a computer.



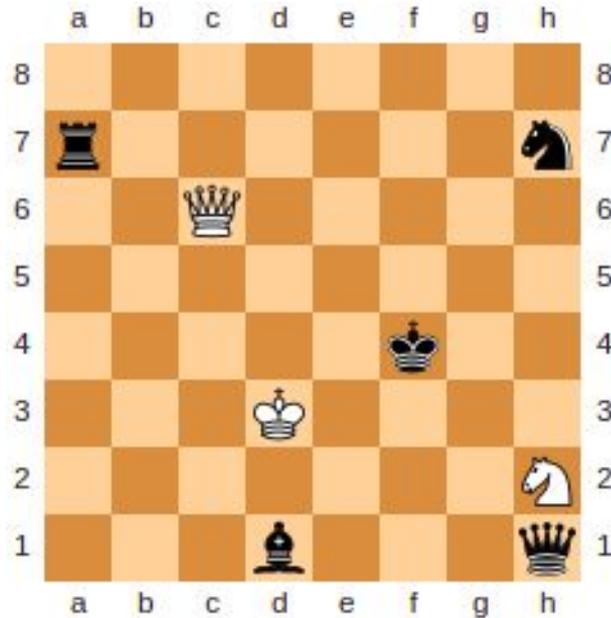
# Endgame Tablebases

- All positions up to 7 pieces (including 2 kings) have been completely analysed and can be perfectly played.
- Research to generate all 8 piece positions are currently underway.
- Once the computer reaches a 7 piece endgame (or fewer pieces in many cases), it consults an endgame tablebase and plays a perfect game.
- The endgame tablebases has opened new understanding of endgame theory.

## LESSON:

- You need to **STUDY** the various techniques for correct endgame play and become skilled in various material setups.

Forced mate in 546... white to move



Number of pieces	Number of positions	Database name	size
2	462	Syzygy	(included in the 5-piece tablebase)
3	368,079	Syzygy	(included in the 5-piece tablebase)
4	125,246,598	Syzygy	(included in the 5-piece tablebase)
5	25,912,594,054	Syzygy	939 MB
6	3,787,154,440,416	Syzygy	150.2 GB
7	423,836,835,667,331	Syzygy	17 TB
7	423,836,835,667,331	Lomonosov	140 TB
8	38,176,306,877,748,245	n/a	1 <a href="#">PB</a>

# Evaluation Function

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# What is an evaluation function?

- Value that expresses the strength of a position.
- Usually values larger than 0.0 indicate a superior position for white and values smaller than 0.0 indicate a superior position for black.
- Usually normalised such that the value for a pawn is about 1.0.
- A computer needs to compute an evaluation for each position it considers - it precisely expresses the evaluation of the position.
- The goal would be to optimize the evaluation function for the player over time.
- Usually, white starts with an advantage of about 0.1 to 0.2
- 0.5 - slight advantage, >1 - advantage, >3 - decisive advantage



# Elements that go into the evaluation function

- Piece value
- Piece square bonuses
- Material imbalance
- Pawn structure (isolated, backward, doubled, connected, connected bonus)
- Piece bonus (penalties to pieces of given colour and type - knight outposts, bishop outposts, trapped rooks, long diagonal bishop, weak queen, etc.)
- Mobility
- Threats (hanging, overload, king threat, pawn push threat, rank threat, weak unopposed pawn, etc.)
- Passed pawn
- Space
- King attack (king danger, shelter strength, pawnless flank, etc)
- Initiative



# Lessons from the evaluation function

- You need to develop the ability to ACCURATELY access the position.
- If you do not consider ALL aspects of the position, you may completely miss some critical contributions to the evaluation of the position.

# Building a search tree

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# Creating a search tree

- A computer needs to construct a search tree of the possible moves from the current position.
- It uses the evaluation function to determine a score associated with each of the generated positions.
- It chooses its moves based on a maximization of the evaluation function.



# Branching Factor

- How many moves can you make in any given position?
- In the beginning, there are 20.
- The average over the duration of a chess game, is about 35.
- This is known as the branching factor.
- After two “plies” (a move by you and your opponent), there can be on average  $35 \times 35 = 1225$  possible positions.
- After only 5 moves by you and your opponent, there are  $35^5 = 2758547353515625$  possible positions...



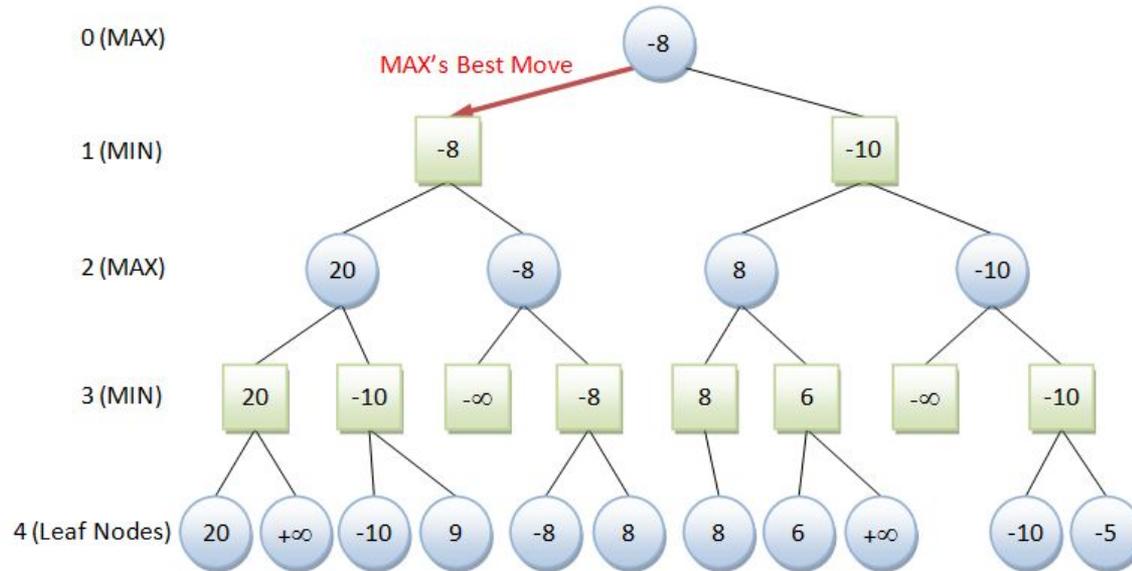
# Reducing the branching factor

- It is impossible to consider all the possible variations.
- To reduce complexity, only a few branches (2 to 4) are considered at each position.
- The computer uses heuristics (possible best move first) to determine which of these to consider first.
- Practice has shown that it is better to analyse deeper, but fewer moves at each point.

## LESSON:

- You need to **TAKE YOUR TIME** to at least consider some continuations for every position.
- You need to **IDENTIFY CANDIDATE MOVES** for further consideration.

# The MiniMax Algorithm

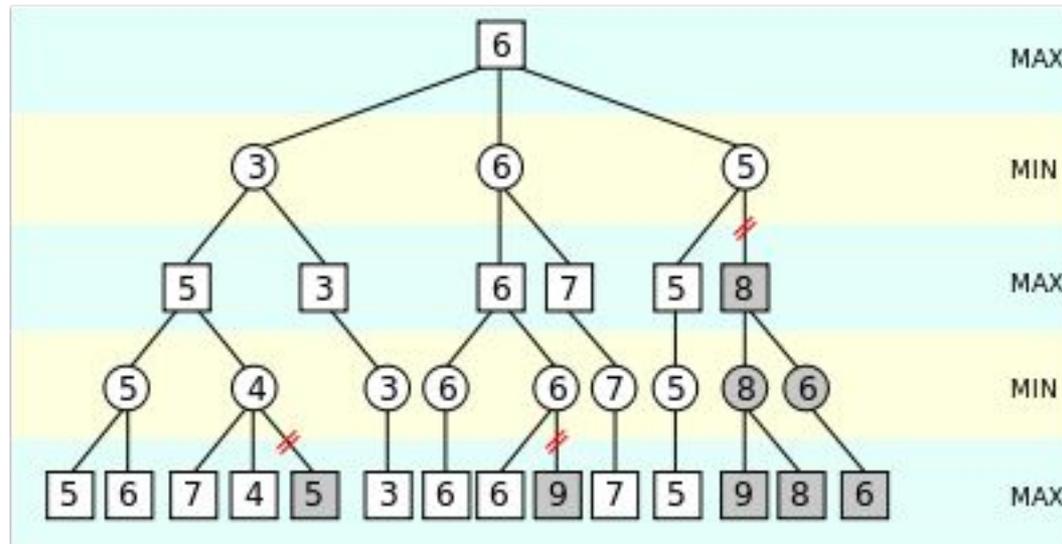




# Lessons from the MiniMax algorithm

- You should always ASSUME that your opponent will play at their BEST.
- Choose moves that are GUARANTEED to give your position the best possible evaluation, not moves where your opponent has potential to optimize their evaluation.
- Consider each POSITION ONCE.

# Alpha Beta Pruning





# Lessons from Alpha-Beta pruning

- Once you find a REFUTATION of one of the moves you consider, you should immediately stop the evaluation of the line, and backtrack the line to a prior position.
- Attempt to CUT OUT unnecessary thinking as soon as possible.



# The NULL move heuristic

- A good heuristic for when the evaluation of a position does not significantly change, is the NULL move heuristic.
- If a player can forfeit a move and his position does not significantly weaken (the evaluation of the position does not change much), then the line need not be further considered.

## LESSONS:

- Sometimes it is worth checking the strength of your position by considering what would happen if your opponent could make two moves consecutively.
- Sometimes you need to stop to consider further moves when your candidate move is good enough.

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Thank You

