



# Chessmaster

A new way to play chess

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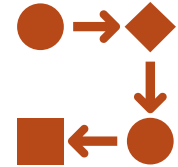
Motivation



State of the Art



Algorithm



Procedure

# Motivation

# Main Goal

- Create an powerful AI, that is capable to beat the best chessplayers in the world
- Provide a platform to train against a strong opponent
- Give the „real“ chessplayers the opportunity to step up their gametactics



# State of the Art

# IBM - Deep Blue

- First real chess engine to beat a reign world champion
- Developed by IBM in 1989
- It implemented the alpha beta search on very large-scale integration but in a brut-force method
  - Developers even denied it being an AI



# Stockfish

- Open source UCI chess engine
  - UCI = Universal Chess interface
- As of 2018 it has become the world strongest chess entity not relying on AI
- It uses Alpha-Beta-Search and Bitboards
  - Alpha-Beta-Search is a optimized version of the MiniMax Algorithm
  - A bitboard is a specialized bit array data structure where each bit corresponds to a game board space or piece.



# DeepMind - AlphaZero

- Uses a Deep Neural Network in combination with a reinforcement learning algorithm
  - Deep Neural Networks are Neural Networks with two or more hidden layers
- The trained network is used to guide the Monte-Carlo Tree Search Algorithm to select the most promising moves
- In the beginning the algorithm does not know anything about the game besides the basic rules
  - After 4 hours of training it was able to beat Stockfish
- Besides Chess AlphaZero also learned Shogi and Go





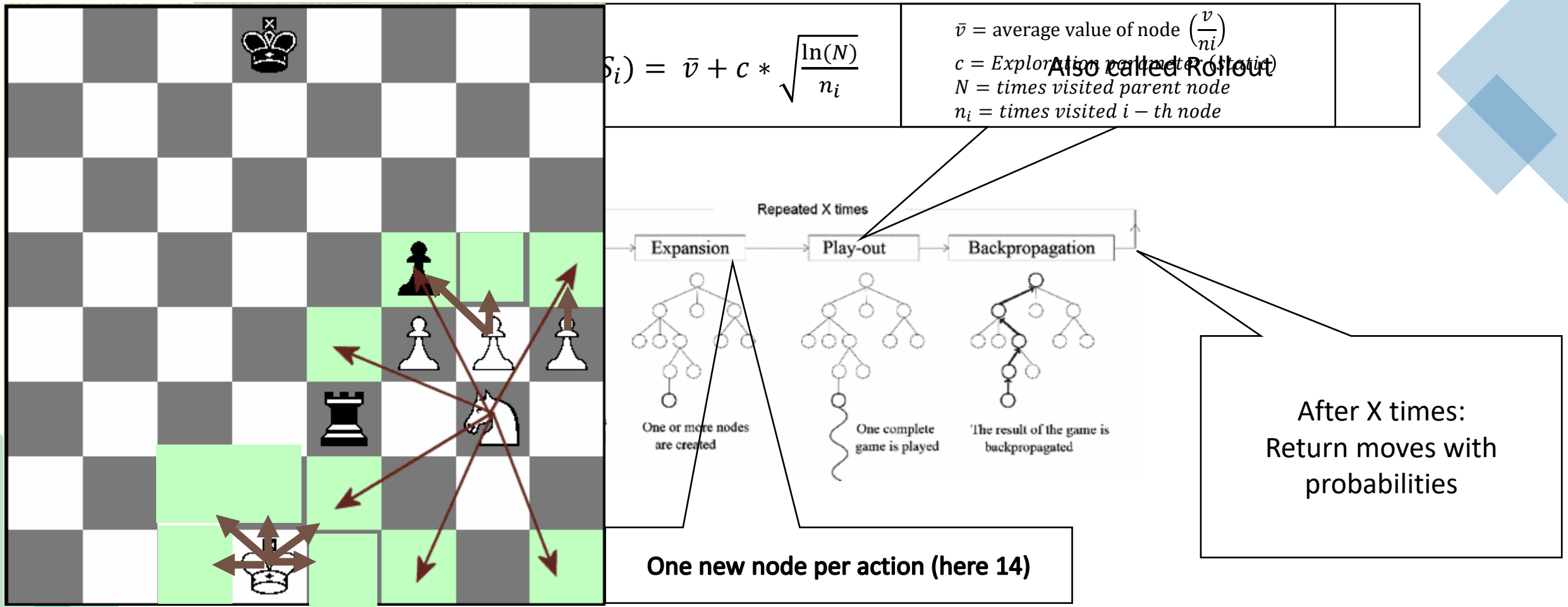


Algorithm explained

# Monte Carlo Tree Search

- Statistical Algorithm
- Described in 2006 by Rémi Coulom
- Used by various AIs, like AlphaZero or MuZero (by Deepmind)
- Improved Efficiency above (true) Tree Search (Brute Force)
- Prevents a large Exploration to Exploitation Trade off!
- Used to find optimal moves

# Monte Carlo Tree Search



# V-Network | Temporal Difference Learning

- V = Value
- V-Network
  - Neural Network (Chess agent)
  - Should output a policy  $\pi$  (to win the game)
- Temporal Difference Learning
  - Model-Free Reinforcement Learning Algorithm
    - Learn from experience by incomplete episodes
  - $V(S_t) = V(S_t) + \alpha * (G_t - V(S_t))$ 
    - $\alpha$  = Learning Rate  $\in [0,1]$
    - $G_t$  = Estimated return
  - Simplest TD-Algorithm (TD(0))
    - $G_t = R_{t+1} + \gamma * V(S_{t+1})$
    - $V(S_t) = V(S_t) + \alpha * (R_{t+1} + \gamma * V(S_{t+1}) - V(S_t))$ 
      - $R_t$  = Reward
    - $\gamma$  = discount factor  $\in [0,1]$

$$R_{t+1} + \gamma * V(S_{t+1}) = \text{TD Target}$$

$$R_{t+1} + \gamma * V(S_{t+1}) - V(S_t) = \text{TD error}$$

# Procedure

# Next Steps

- Comparison of the evaluation possibilities of the algorithm
- Human versus computer (Graphical user interface)
- Summarize scientific papers
- Preparation of the elaboration

# Evaluation of Algorithm

- Elo rating system (Comparison of two chess players)
- Evaluation of the chess game
  - Evaluation of Position
  - Material on the board
  - Pieces Activity



Any Questions?





Thank you for your  
Attention