

# Monte Carlo II

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# Monte Carlo Logic

Question: In Monte Carlo, what is used to measure what path is taken?

# Monte Carlo Logic

$$UCT = Q(v) + E * \sqrt{(\ln(N(v)) / N(v\_child))}$$

UCT stands for **U**pper **C**onfidence bound for **T**rees. It is a metric that takes into account:

Q(v): Our estimated value of the current node

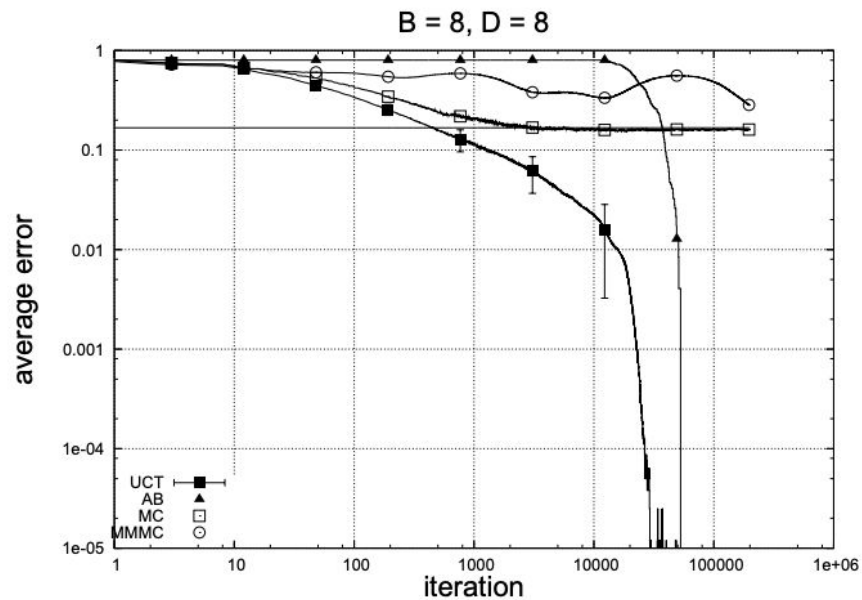
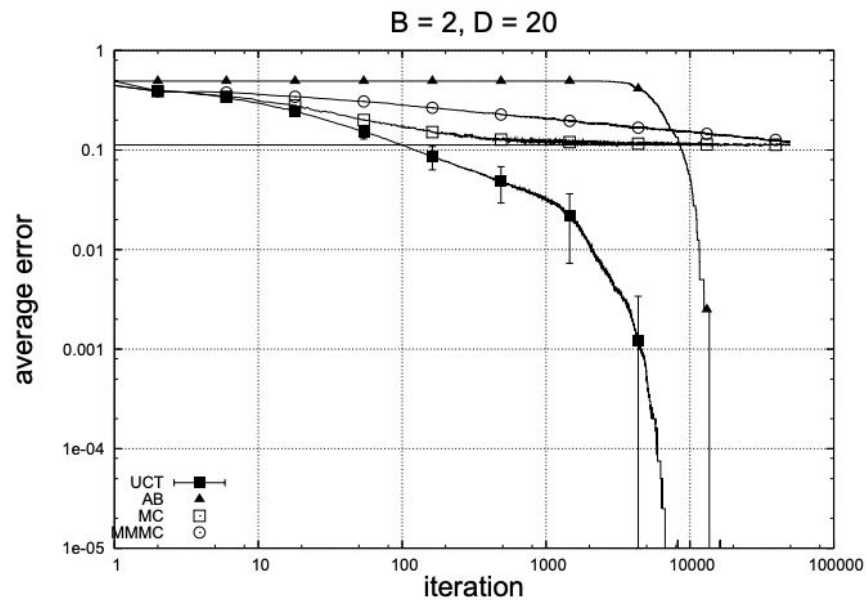
E: Our epsilon, or exploration value

N(v): How many times the current node was visited

N(v\_child): How many times the child was visited

It was originally proposed by Levente Kocsis and Csaba Szepesvari at the Hungarian Academy of Sciences in 2006. What they found was that UCT was able to work much quicker with less data and achieve the same error rate as other algorithms.

(AB: Alpha-Beta, MC: Vanilla MC, MMMC: Minimax Value Update)



**Fig. 2.** Failure rate in P-games. The 95% confidence intervals are also shown for UCT.

# UCT

## Benefits

- Allows for further exploration of nodes by including number of visits
- Provides a flexible, general-use method for analyzing a given environment

## Limitations

- Reliance on Exploration can hinder learning

# UCT Improvements

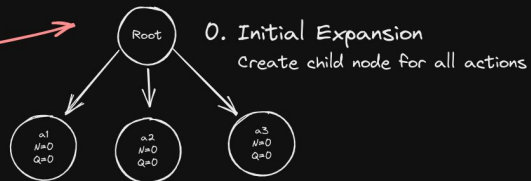
## RAVE

- Improves data collection during MCTS by keeping more information on the actions available regardless if they were taken
- Allows for MCTS to learn quicker earlier on when data is scarce
- Implemented in the backpropagation function

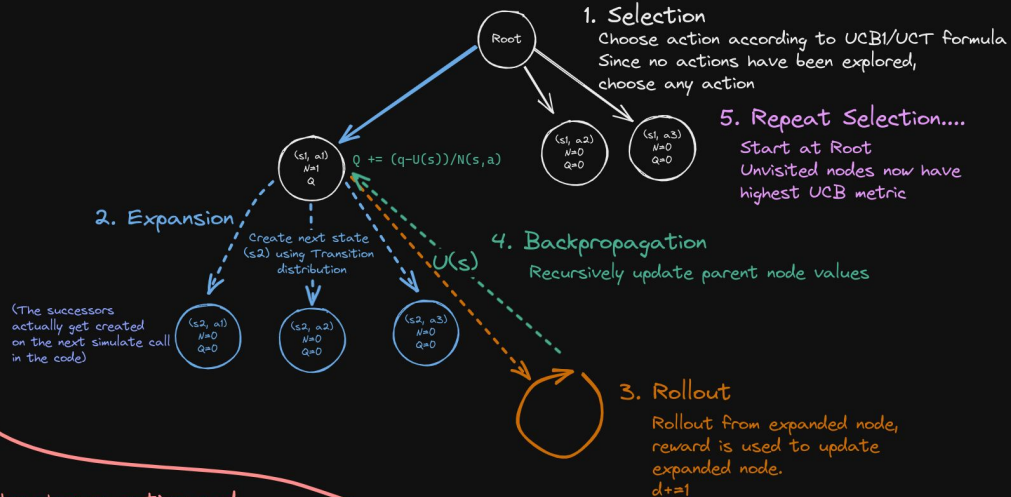
## PUCT

- Generates weights for each initial action in order to better predict which action will be optimal.
-

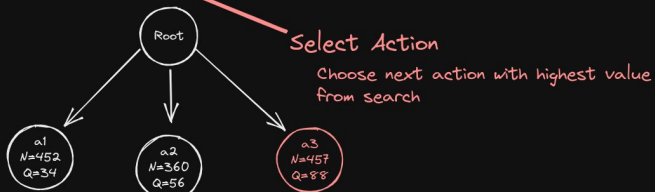
While compute available...



For  $d = 1$  : max depth...



Take chosen action and transition to next state  
Repeat search





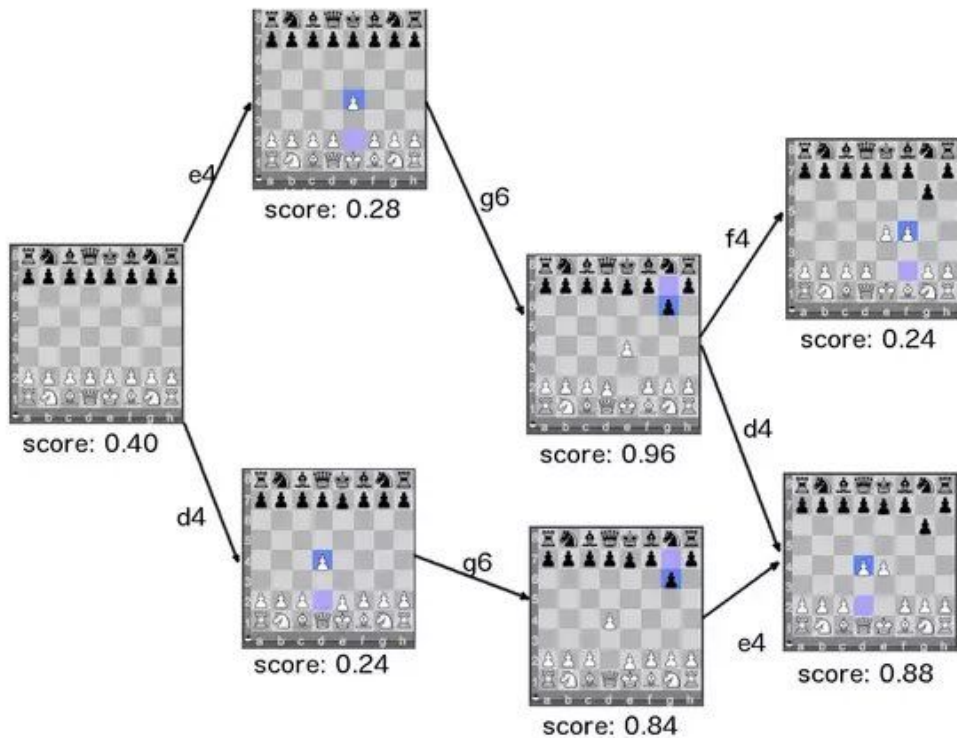
# Case Study I: Chess



Q1: Just looking at this board, what would 100 be a reference to?

Q2: In a game of chess, how would you assess the effectiveness of a board?

# Case Study I: Chess



# Case Study I: Chess

- Terminal State:
  - King has no moves left and is in check
  - no actions are possible
- Actions
  - Moving a pawn
- Rewards
  - 1 if win
  - -1 if lose
  - 0 if draw
  - Generates legal moves only so no need to worry about punishing illegal moves



# Case Study II: Tic Tac Toe

<https://ai-boson.github.io/mcts/>

# Required Functions for Running MCTS

- Node representation that can store
  - The parent node
  - Action taken from the parent
  - How many times it was visited
  - The rewards from its children
  - List of Actions from that node that have not been tried
  - The children possible from the node (future states)
- As actions are taken, new nodes are generated.
- All MCTS actions take place in this node class

# Further Reading

- AlphaGo and AlphaZero papers
- “A Survey of MCTS Methods” published in IEEE’s Transactions on Computational Intelligence and AI in Games, Vol. 4, No. 1, March 2012

<http://www.incompleteideas.net/609%20dropbox/other%20readings%20and%20resources/MCTS-survey.pdf>

# Works Cited

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K. Matsuzaki, “Empirical analysis of PUCT algorithm with evaluation functions of different quality,” *2018 Conference on Technologies and Applications of Artificial Intelligence (TAAI)*, 2018. doi:10.1109/taai.2018.00043

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